Balloon Hashing: a Provably Memory-Hard Function with a Data-Independent Access Pattern

Dan Boneh¹, Henry Corrigan-Gibbs¹, and Stuart Schechter²

¹ Stanford University
² Microsoft Research

Abstract. We present the Balloon algorithm for password hashing. This is the first cryptographic hash function that: (i) has proven memory-hardness properties in the random-oracle model, (ii) uses a password-independent access pattern, and (iii) meets—and often exceeds—the performance of the best heuristically secure password-hashing algorithms. Memory-hard functions require a large amount of working space to evaluate efficiently and, when used for password hashing, they dramatically increase the cost of offline dictionary attacks. In this work, we leverage a previously unstudied property of a certain class of graphs ("random sandwich graphs") to analyze the memory-hardness of the Balloon algorithm. The techniques we develop are general: we also use them to give a proof of security of the scrypt and Argon2i password-hashing functions, in the random-oracle model. To motivate the need for security proofs in the area, we demonstrate a practical attack against Argon2i that successfully evaluates the function with less space than was previously claimed possible. Finally, we discuss recent important work on parallel attacks against memory-hard functions with password-independent access patterns, and we propose a defense against them. We experiment with the Balloon hashing algorithm and report on its performance relative to other claimed memory-hard functions.

Keywords: memory-hard functions, password hashing, pebbling arguments, time-space trade-offs, sandwich graph, Argon2, scrypt.

1 Introduction

The staggering number of institutions that have had their authentication databases or password files breached in recent months demonstrates the importance of cryptographic protection for stored passwords. In 2015 alone, attackers stole files containing users’ login names, password hashes, and contact information from many large and well-resourced organizations, including LastPass [86], Harvard [50], E*Trade [67], ICANN [48], Costco [44], T-Mobile [83], the University of Virginia [81], and a large number of others [71]. In this environment, systems administrators must operate under the assumption that attackers will eventually gain access to sensitive authentication information, such as password hashes.
and salts, stored on their computer systems. After a compromise, the secrecy of
user passwords rests on the cost to an attacker of mounting an offline dictionary
attack against the stolen file of hashed passwords.

An ideal password-hashing function has the property that it costs as much
for an attacker to compute the function as it does for the legitimate authentica-
tion server to compute it. Standard cryptographic hashes completely fail in this
regard: it costs 100 000× more to compute a SHA-256 hash on a general-purpose
x86 CPU (as an authentication server would use) than it does to compute on
special-purpose hardware (such as the ASICs that an attacker would use) [20].
Iterating a standard cryptographic hash function, as is done in bcrypt [72] and
PBKDF2 [46], increases the absolute cost to the attacker and defender, but the
attacker’s 100 000× relative cost advantage remains.

Memory-hard functions help close the efficiency gap between the attacker
and defender in the setting of password hashing [6, 17, 39, 61, 65]. Memory-hard
functions exploit the observation that on-chip memory is just as costly to power
on special-purpose hardware as it is on a general-purpose CPU. If evaluating the
password-hashing function requires large amounts of memory, then an attacker
using special-purpose hardware has little cost advantage over the legitimate au-
thentication server (using a standard x86 machine, for example) at running the
password-hashing computation. Memory consumes a large amount of on-chip
area, so the high memory requirement ensures that a special-purpose chip can
only compute a few instances of the function in parallel.

Optimal memory-hard functions, with security parameter \(n\), have a space-
time product that satisfies \(S \cdot T \in \Omega(n^2)\), irrespective of the strategy used to
compute the function. The challenge is to construct a function that provably
satisfies this bound with the largest possible constant multiple on the \(n^2\) term.

In this paper, we introduce the Balloon memory-hard function for password
hashing. This is the first password-hashing function to simultaneously satisfy
three important design goals [61]:

- **Proven memory-hard.** We prove, in the random-oracle model [12], that com-
  puting the Balloon function with space \(S\) and time \(T\) requires \(S \cdot T \geq n^2 / 8\)
  (approximately). As the adversary’s space usage decreases, we prove even
  sharper time-space lower bounds.

  To motivate our interest in memory-hardness proofs, we demonstrate in Sec-
  tion 5 an attack against the Argon2i password hashing function [17], winner
  of a recent password-hashing design competition [61]. The attack evaluates
  the function with far less space than claimed without changing the time re-
  quired to compute the function. We also give a proof of security for Argon2i,
  in the random-oracle model, that shows that significantly more powerful
  attacks are impossible.

- **Password-independent memory access pattern.** The memory access pattern
  of the Balloon algorithm is independent of the password being hashed. As we
  describe in Section 4.2, password-hashing functions that lack this property
  are vulnerable to a crippling attack in the face of an adversary who learns
the memory access patterns of the hashing computation (e.g., via cache side-channels [22, 59, 84]).

- Performant. The Balloon algorithm is easy to implement and it matches or exceeds the performance of the fastest comparable password-hashing algorithms, Argon2i [17] and Catena [39], when instantiated with standard cryptographic primitives (Section 6).

We analyze the memory-hardness properties of the Balloon function using pebble games, which are arguments about the structure of the data-dependency graph of the underlying computation [51, 62, 64, 78, 82]. Our analysis uses the framework of Dwork, Naor, and Wee [34]—later applied in a number of cryptographic works [4, 6, 35, 36, 39]—to relate the hardness of pebble games to the hardness of certain computations in the random-oracle model [12].

The crux of our analysis is a new observation about the properties of “random sandwich graphs,” a class of graphs studied in prior work on pebbling [4, 6]. To show that our techniques are broadly applicable, we apply them in Appendices F and G to give simple proofs of memory-hardness, in the random-oracle model, for the Argon2i and scrypt functions. We prove stronger memory-hardness results about the Balloon algorithm, but these auxiliary results about Argon2i and scrypt may be of independent interest to the community.

The performance of the Balloon Hashing algorithm is surprisingly good, given that our algorithm offers stronger proven security properties than other practical memory-hard functions with password-independent access patterns. For example, if we configure Balloon to use Blake2b as the underlying hash function [8], run the construction for five “rounds” of hashing, and set the space parameter to require the attacker to use 1 MiB of working space to compute the function, then we can compute Balloon Hashes at the rate of 13 hashes per second on a modern server, compared with 12.8 for Argon2i, and 2.1 for Catena DBG (when Argon2i and Catena DBG are instantiated with Blake2b as the underlying cryptographic hash function).

3 Caveat: Multiple-Instance Security. The definition of memory-hardness we use puts a lower-bound on the time-space product of computing a single instance of the Balloon function. In reality, an adversary mounting a dictionary attack would want to compute billions of instances of the Balloon function, perhaps using many processors running in parallel. Alwen and Serbinenko [6], formalizing earlier work by Percival [65], introduce a new computational model—the parallel random-oracle model (pROM)—and a memory-hardness criterion that addresses the shortcomings of the traditional model. In recent work, Alwen and Blocki prove the surprising result that no function that uses a password-independent access pattern can be optimally memory-hard in the pROM [3]. In addition, they give a special-purpose pROM algorithm for computing Argon2i, Balloon, and other practical (single-instance) memory-hard functions with some space savings. We discuss this important issue and the relevant related work in Section 4.

3 The relatively poor performance of Argon2i here is due to the attack we present in Section 5. It allows an attacker to save space in computing Argon2i with no increase in computation time.
Contributions. In this paper, we

- introduce and analyze the Balloon functions, which have stronger provable security guarantees than all prior practical memory-hard functions (Section 3),
- explain how to ameliorate the danger of massively parallel attacks against memory-hard functions with password-independent access patterns (Section 4)
- present a practical memory-saving attack against the Argon2i password-hashing algorithm (Section 5), and
- prove the first known time-space lower bounds for Argon2i (Appendix F) and an idealized variant of scrypt (Appendix G), in the random-oracle model.

Overall, we show that password hashing is an area where proofs of memory-hardness are vitally important and can be obtained for practical constructions. In particular, memory-hard functions such as the Balloon function can be at once practical and provably secure.

Notation. Throughout this paper, Greek symbols ($\alpha, \beta, \gamma, \lambda$, etc.) typically denote constants greater than one. We use $\log_2(\cdot)$ to denote a base-two logarithm and $\log(\cdot)$ to denote a logarithm when the base is not important. For a finite set $S$, the notation $x \leftarrow R S$ indicates sampling an element of $S$ uniformly at random and assigning it to the variable $x$.

2 Security Definitions

This section summarizes the high-level security and functionality goals of a password hashing function in general and the Balloon hashing algorithm in particular. We draw these aims from prior work on password hashing [65, 72] and also from the requirements of the recent Password Hashing Competition [61].

2.1 Syntax

The Balloon password hashing algorithm takes four inputs: a password, salt, time parameter, and space parameter. The output is a bitstring of fixed length (e.g., 256 or 512 bits). The password and salt are standard [56], but we elaborate on the role of the latter parameters below.

Space Parameter (Buffer Size). The space parameter, which we denote as “$n$” throughout, indicates how many fixed-size blocks of working space the hash function will require during the course of its computation, as in scrypt [65]. At a high level, a memory-hard function should be “easy” to compute with $n$ blocks of working space and should be “hard” to compute with much less space than that. We make this notion precise later on.

Time Parameter (Number of Rounds). The Balloon function takes as input a parameter $r$ that determines the number of “rounds” of computation it performs.
As in bcrypt [72], the larger the time parameter, the longer the hash computation will take. As computational power increases, users can increase this time parameter to keep the number of wall-clock seconds required to compute each hash near-constant. In addition, the choice of $r$ has an effect on the memory-hardness properties of the scheme: the larger $r$ is, the longer it takes to compute the function in small space.

2.2 Memory-Hardness

We say that a function $f_n$ on space parameter $n$ is (single-instance) memory-hard in the random-oracle model [12] if, for all adversaries computing $f_n$ with high probability using space $S$ and $T$ random oracle queries, we have that

$$S \cdot T \in \Omega(n^2).$$

This definition deserves a bit of elaboration. Following Dziembowski et al. [36] we say that an algorithm “uses space $S$” if the entire configuration of the Turing Machine (or RAM machine) computing the algorithm requires at least $S$ bits to describe. When, we say that an algorithm computes a function “with high probability,” we mean that the probability that the algorithm computes the function is non-negligible as the output size of the random oracle and the space parameter $n$ tend to infinity. In practice, we care about the adversary’s concrete success probability, so we avoid asymptotic notions of security wherever possible. In addition, as we discuss in the evaluation section (Section 6), the exact value of the constant hidden inside the $\Omega(\cdot)$ is important for practical purposes, so our analysis makes explicit and optimizes these constants.

A function that is memory-hard under this definition requires the adversary to use either a lot of working space or a lot of execution time to compute the function. Functions that are memory-hard in this way are not amenable to implementation in special-purpose hardware (ASIC), since the cost to power a unit of memory for a unit of time on an ASIC is the same as the cost on a commodity server. An important limitation of this definition is that it does not take into account parallel attacks or multiple-instance attacks, which we discuss in Section 4.

2.3 Password-Independent Access Pattern

A first-class design goal of the Balloon algorithm is to have a memory access pattern that is independent of the password being hashed. (We allow the data-access pattern to depend on the salt, since the salts can be public.) As mentioned above, employing a password-independent access pattern reduces the risk that information about the password will leak to other users on the same machine via cache or other side-channels [22, 59, 84]. This may be especially important in cloud-computing environments, in which many mutually distrustful users share a single physical host [75].
Creating a memory-hard function with a password-independent access pattern presents a technical challenge: since the data-access pattern depends only upon the salt—which an adversary who steals the password file knows—the adversary can compute the entire access pattern in advance of a password-guessing attack. With the access pattern in hand, the adversary can expend a huge amount of effort to find an efficient strategy for computing the hash function in small space. Although this pre-computation might be expensive, the adversary can amortize its cost over billions of subsequent hash evaluations. A function that is memory-hard and that uses a password-independent data access pattern must be impervious to all small-space strategies for computing the function so that it maintains its strength in the face of these pre-computation attacks. (Indeed, as we discuss in Section 4, Alwen and Blocki show that in some models of computation, memory-hard functions with password-independent access patterns do not exist [3].)

2.4 Collision Resistance, etc.

If necessary, we can modify the Balloon functions so that they provide the standard properties of second-preimage resistance and collision resistance [54]. It is possible to achieve these properties in a straightforward way by composing the Balloon function $B$ with a standard cryptographic hash function $H$ as

$$H_B(\text{passwd}, \text{salt}) := H(\text{passwd}, \text{salt}, B(\text{passwd}, \text{salt})).$$

Now, for example, if $H$ is collision-resistant, then $H_B$ must also be. That is because any inputs $(x_p, x_s) \neq (y_p, y_s)$ to $H_B$ that cause $H_B(x_p, x_s) = H_B(y_p, y_s)$ immediately yield a collision for $H$ as:

$$(x_p, x_s, B(x_p, x_s)) \quad \text{and} \quad (y_p, y_s, B(y_p, y_s)),$$

no matter how the Balloon function $B$ behaves.

3 Balloon Hashing Algorithm

In this section, we present the Balloon hashing algorithm.

3.1 Algorithm

The algorithm uses a standard (non-memory-hard) cryptographic hash function $H : \mathbb{Z}_N \times \{0, 1\}^{2k} \to \{0, 1\}^k$ as a subroutine, for some large integer $N$. For the purposes of our analysis, we model the function $H$ as a random oracle [12].

The Balloon algorithm uses a large memory buffer as working space and we divide this buffer into contiguous blocks. The size of each block is equal to the

---

4 We are eliding important definitional questions about what it even means, in a formal sense, for a function to be collision resistant [15, 76].
output size of the hash function $H$. Our analysis is agnostic to the choice of hash function, except that, to prevent issues described in Appendix B.3, the internal state size of $H$ must be at least as large as its output size. Since $H$ maps blocks of $2k$ bits down to blocks of $k$ bits, we sometimes refer to $H$ as a cryptographic compression function.

The Balloon function operates in three steps (Figure 1):

1. **Expand.** In the first step, the Balloon algorithm fills up a large buffer with pseudo-random bytes derived from the password and salt by repeatedly invoking the compression function $H$ on a function of the password and salt.

2. **Mix.** In the second step, the Balloon algorithm performs a “mixing” operation $r$ times on the pseudo-random bytes in the memory buffer. The user-specified round parameter $r$ determines how many rounds of mixing take place. At each mixing step, for each block $i$ in the buffer, the routine updates the contents of block $i$ to be equal to the hash of block $(i - 1) \mod n$, block $i$, and $\delta$ other blocks chosen “at random” from the buffer. (See Theorem 1 for an illustration of how the choice of $\delta$ affects the security of the scheme.)
Since the Balloon functions are deterministic functions of their arguments, the dependencies are not chosen truly at random but are sampled using a pseudorandom stream of bits generated from the user-specific salt.

3. **Extract.** In the last step, the Balloon algorithm just outputs the last block of the buffer.

**Multi-Core Machines.** A limitation of the Balloon algorithm as described is that it does not allow even limited parallelism, since the value of the $i$th block computed always depends on the value of the $(i-1)$th block. To increase the rate at which the Balloon algorithm can fill memory on a multi-core machine with $M$ cores, we can define a function that invokes the Balloon function $M$ times in parallel and XORs all the outputs. If $\text{Balloon}(p, s)$ denotes the Balloon function on password $p$ and salt $s$, then we can define an $M$-core variant $\text{Balloon}_M(p, s)$ as:

$$\text{Balloon}_M(p, s) := \text{Balloon}(p, s||"1") \oplus \cdots \oplus \text{Balloon}(p, s||"M").$$

A straightforward argument shows that computing this function requires computing $M$ instances of the single-core Balloon function. Existing password hashing functions deploy similar techniques on multi-core platforms [17, 39, 65, 66].

3.2 **Main Security Theorem**

The following theorem demonstrates that attackers who attempt to compute the Balloon function in small space must pay a large penalty in computation time. The complete theorem statement is given in Theorem 33 (Appendix E).

**Theorem 1 (informal)** Let $A$ be an algorithm that computes the $n$-block $r$-round Balloon function with security parameter $\delta \geq 3$, where $H$ is modeled as a random oracle. If $A$ uses at most $S$ blocks of buffer space then, with overwhelming probability, $A$ must run for time (approximately) $T$, such that

$$S \cdot T \geq \frac{r \cdot n^2}{8}.$$

Moreover, under the stated conditions, one obtains the stronger bound:

$$S \cdot T \geq \frac{(2^r - 1)n^2}{8}$$

if

$$\begin{cases} \delta = 3 \text{ and } S < n/64 \text{ or}, \\ \delta = 4 \text{ and } S < n/32 \text{ or}, \\ \delta = 5 \text{ and } S < n/16 \text{ or}, \\ \delta = 7 \text{ and } S < n/8. \end{cases}$$

The theorem shows that, when the adversary’s space usage falls below a certain threshold (parameterized by $\delta$), the computation time increases exponentially in the number of rounds $r$. For example, when $\delta = 7$ and the space $S$ is less than $n/8$, the time to evaluate Balloon is at least $2^r$ times the time to evaluate it with space $n$. Thus, attackers who attempt to compute the Balloon function in very small space must pay a large penalty in computation time.
The proof of the theorem is given in the appendices: Appendix B introduces graph pebbling arguments, which are the basis for our memory-hardness proofs. Appendix C proves a key combinatorial lemma required for analysis of the Balloon functions. Appendix D recalls sandwich graphs and proves that stacks of random sandwich graphs are hard to pebble. Appendix E puts all of the pieces together to complete the proof of Theorem 1.

Here we sketch the main ideas in the proof of Theorem 1.

\[
\begin{align*}
v_1 &= \text{hash(input, "0")} \\
v_2 &= \text{hash(input, "1")} \\
v_3 &= \text{hash}(v_1, v_2) \\
v_4 &= \text{hash}(v_2, v_3) \\
v_5 &= \text{hash}(v_3, v_4) \\
\text{return } v_5
\end{align*}
\]

Fig. 2: An example computation (left) and its corresponding data-dependency graph (right).

**Proof idea.** The proof makes use of pebbling arguments, a classic technique for analyzing computational time-space trade-offs \([45, 51, 64, 68, 78, 85]\) and memory-hard functions \([6, 34, 35, 39]\). We apply pebbling arguments to the data-dependency graph corresponding to the computation of the Balloon function (See Figure 2 for an example graph). The graph contains a vertex for every random oracle query made during the computation of Balloon: vertex \(v_i\) in the graph represents the response to the \(i\)th random-oracle query. An edge \((v_i, v_j)\) indicates that the input to the \(j\)th random-oracle query depends on the response of the \(i\)th random-oracle query.

The data-dependency graph for a Balloon computation naturally separates into \(r + 1\) layers—one for each round of mixing (Figure 3). That is, a vertex on level \(\ell \in \{1, \ldots, r\}\) of the graph represents the output of a random-oracle query made during the \(\ell\)th mixing round.

The first step in the proof shows that the data-dependency graph of a Balloon computation satisfies certain connectivity properties, defined below, with high probability. The probability is taken over the choice of random oracle \(H\), which determines the data-dependency graph.

Consider placing a pebble on each of a subset of the vertices of the data-dependency graph of a Balloon computation. Then, as long as there are “not too many” pebbles on the graph, we show that the following two properties hold w.h.p.:

- **Well-Spreadedness.** For every set of \(k\) consecutive vertices on some level of the graph, at least a quarter of the vertices on the prior level of the graph are on unpebbled paths to these \(k\) vertices (Lemma 28).
Fig. 3: The Balloon data-dependency graph on $n = 8$ blocks and $r = 2$ rounds, drawn with $\delta = 1$ for simplicity. (The real construction uses $\delta \geq 3$.) The dashed edges are fixed and the solid edges are chosen pseudorandomly by applying the random oracle to the salt.

Expansion. All sets of $k$ vertices on any level of the graph have unpebbled paths back to at least $2k$ vertices on the prior level (Lemma 29). The value of $k$ depends on the choice of the parameter $\delta$.

The next step is to show that every graph-respecting algorithm computing the Balloon function requires large space or time. We say that an adversary $A$ is graph respecting if for every $i$, adversary $A$ makes query number $i$ to the random-oracle only after it has in storage all of the values that this query takes as input.

We show, using the well-spreadedness and expansion properties of the Balloon data-dependency graph, that every graph-respecting adversary $A$ must use space $S$ and time $T$ satisfying $S \cdot T \geq n^2/8$, w.h.p. over the choice of $H$ (Lemma 31). We use the graph structure in the proof as follows: fix a set of $k$ values that the adversary has not yet computed. Then the graph properties imply that these $k$ values have many dependencies that a space-$S$ adversary cannot have in storage. Thus, making progress towards computing the Balloon function in small space requires the adversary to undertake a huge amount of recomputation.

The final step uses a technique of Dwork, Naor, and Wee [34]. They use the notion of a graph labeling to convert a directed acyclic graph $G$ into a function $f_G$ (Definition 3). They prove that if $G$ is a graph that is infeasible for time-$T$ space-$S$ graph-respecting pebbling adversaries to compute, then it is infeasible for time-$T'$ space-$S'$ arbitrary adversaries to compute the labeling function $f_G$, with high probability in the random-oracle model (Theorem 4), where $T' \approx T$ and $S' \approx S$.

---

5 This description is intentionally informal—see Appendix B for the precise statement.
We observe that Balloon computes the function $f_G$ where $G$ is the Balloon data-dependency graph (Claim 32). We then directly apply the technique of Dwork, Naor, and Wee to obtain an upper bound on the probability that an arbitrary adversary can compute the Balloon function in small time and space (Theorem 33).

4 Memory Hardness Under Parallel Attacks

The Balloon Hashing algorithm achieves the notion of memory-hardness introduced in Section 2.2: an algorithm for computing Balloon must, with high probability in the random-oracle model, use (roughly) time $T$ and space $S$ that satisfy $S \cdot T \in \Omega(n^2)$. Using the time-space product in this way as a proxy metric for computation cost is natural, since it approximates the area-time product required to compute the function in hardware [65].

4.1 Modeling a Parallel Attacker

As Alwen and Serbinenko [6] point out, there are two key limitations to the standard definition of memory-hardness in which we prove security. First, the definition yields a single-instance notion of security. That is, our definition of memory-hardness puts a lower-bound on the $ST$ cost of computing the Balloon function once, whereas in a password-guessing attack, the adversary potentially wants to compute the Balloon function billions of times. Second, the definition treats a sequential model of computation—in which the adversary can make a single random-oracle query at each time step. In contrast, a password-guessing adversary may have access to thousands of computational cores operating in parallel.

To address the limitations of the conventional single-instance sequential adversary model (which we use for our analysis of the Balloon function), Alwen and Serbinenko introduce a new adversary model and security definition. Essentially, they allow the adversary to make many parallel random-oracle queries at each time step. In this “parallel random-oracle model” (pROM), they attempt to put a lower bound on the sum of the adversary’s space usage over time: $\sum t S_t \in \Omega(n^2)$, where $S_t$ is the number of blocks of space used in the $t$th computation step. We call a function that satisfies this notion of memory-hardness in the pROM an amortized memory-hard function.\footnote{Bellare, Ristenpart, and Tessaro consider a different type of multi-instance security [11]: they are interested in key-derivation functions $f$ with the property that finding $(x_1, \ldots, x_m)$ given $(f(x_1), \ldots, f(x_m))$ is roughly $m$ times as costly as inverting $f$ once. Stebila et al. [80] and Groza and Warinschi [43] investigate a similar multiple-instance notion of security for client puzzles [33] and Garay et al. [40] investigate related notions in the context of multi-party computation.}

\footnote{In the original scrypt paper, Percival [65] also discusses parallel attacks and makes an argument for the security of scrypt in the pROM.}
To phrase the definition in different terms: Alwen and Serbinenko look for functions $f$ such that computing $f$ requires a large amount of working space at many points during the computation of $f$. In contrast, the traditional definition (which we use) proves the weaker statement that the adversary computing $f$ must use a lot of space at some point during the computation of $f$.

An impressive recent line of work has uncovered many new results in this model:

- Alwen and Serbinenko [6] construct a function that is “almost” amortized memory-hard in the pROM (up to polylogarithmic factors) and uses a password-independent access pattern. Unfortunately, their construction is apparently impractical: they prove that the amortized time-space product is roughly $S \cdot T \geq n^2 / (10000 \cdot \log_{10}^2 n)$. Until $n$ is far larger than we would expect in practice, this bound is not meaningful.
- Alwen et al. [4] prove under combinatorial conjectures that scrypt is amortized memory-hard in the pROM. Unlike Balloon, scrypt uses a data-dependent access pattern—which we would like to avoid—and the data-dependence of scrypt’s access pattern seems fundamental to their security analysis.
- Alwen and Blocki [3] show that, in the pROM, there does not exist a perfectly memory-hard function (in the amortized sense) that uses a password-independent memory-access pattern. Additionally, they give special-case pROM algorithms for computing many candidate password-hashing algorithms in the pROM. Their algorithm computes Balloon, for example, using an amortized time-space product of (roughly) $O(n^{7/4})$—instead of the $\Omega(n^2)$ one would hope for.

As far as practical constructions go, these results leave the practitioner with two options, each of which has a downside:

**Option 1.** Use scrypt, which has a proof of security (under conjectures) in the pROM, but which uses a password-dependent access pattern and is weak in the face of an adversary that can learn memory access information. We describe the attack in Section 4.2 below.

**Option 2.** Use Balloon Hashing, which uses a password-independent access pattern and has a proof of security in the random-oracle model, but weakens in the face of a massively parallel attack.

A good practical solution is to hash passwords using a careful composition of Balloon and scrypt, as we explain in Section 4.3. One function defends against memory access pattern leakage and the other defends against massively parallel attacks.

---

8 A recent addendum to the paper suggests that the combinatorial conjectures on which the scrypt proof of security is based may be false [5, Section 0].

9 It would take some additional work to determine the parameter sizes at which the attacks would become practical to implement in hardware today. That said, the fact that such pROM attacks exist at all are absolutely a practical concern.
4.2 Attacking Scrypt Under Access Pattern Leakage

We argue first that Option 1 (using vanilla scrypt) is practically unwise. Why? We show that if an attacker can obtain just a few bits of information about the data-access pattern that scrypt makes when hashing a target user’s password, the attacker can mount a very efficient dictionary attack against that user’s scrypt-hashed password in constant space. To demonstrate this, we recall a folklore attack on scrypt (described by Forler et al. [39]).

Say that the attacker observes the memory access pattern of scrypt as it operates on a target user’s password and that the attacker gets ahold of the target user’s password hash and salt (e.g., by stealing the /etc/shadow file on a Linux system). Say that scrypt, when hashing on the target’s password, accesses the memory blocks at addresses \((A_1, A_2, A_3, \ldots)\). The attacker can now mount a dictionary attack against a stolen password file as follows:

- Guess a candidate password \(p\) to try.
- Run scrypt on the password \(p\) and the salt (which is in the password file), but discard all blocks of data generated except the most recent one. Recompute the values of any blocks needed later on in the computation.
- Say that scrypt on the candidate password accesses block \(A'_1\) first. If \(A_1 \neq A'_1\), terminate the execution and try another candidate password.
- Say that scrypt on the candidate password accesses block \(A'_2\) next. If \(A_2 \neq A'_2\), terminate the execution and try another candidate password.
- Continue in this way until recovering the target’s password...

The probability that a candidate password and the real password agree on the first \(k\) memory addresses is \(n^{-k}\), and the expected time required to run each step of the algorithm is \(n/2\) for \(n\)-block scrypt, since recomputing each discarded memory block takes this much time on average. The expected time to compute one candidate password guess is then:

\[
E[T] = n + \sum_{k=1}^{n} \frac{k(n/2)}{n^k} \leq n \left( 1 + \frac{1}{2} \sum_{k=1}^{n} \frac{k}{n^k} \right) \leq n \left( 1 + \frac{1}{2} n(1/n) \right) \in O(n).
\]

Since the space usage of this attack algorithm is constant, the time-space product is only \(O(n)\) instead of \(\Omega(n^2)\), so scrypt essentially loses its memory-hardness completely in the face of an attacker who learns a few bits of the memory access pattern.\(^{11}\)

Of course, a valid question is whether a real-world attacker could ever learn any bits of the access pattern of scrypt on a user’s password. It is certainly plausible that an attacker could use cache-timing channels—such as exist for certain block ciphers [22]—to extract memory-access pattern information, especially if a malicious user and an honest user are colocated on the same physical machine.

\(^{10}\) See Appendix G for a sketch of the core scrypt algorithm.

\(^{11}\) Here we have assumed that the attacker knows the entire access pattern, but a similarly devastating attack applies even if the attacker only knows the first few memory address locations.
(e.g., as in a shared compute cluster). Whether or not these attacks are practical today, it seems prudent to design our password-hashing algorithms to withstand attacks by the strongest possible adversaries.

4.3 A “Best of Both Worlds” Construction

We have argued that Option 1 above (using vanilla scrypt) is unwise. For the moment, let us stipulate also that the pROM attacks on vanilla Balloon, and all other practical password hashing algorithms using data-independent access patterns, make these algorithms less-than-ideal to use on their own. Can we somehow combine the two constructions to get a “best-of-both-worlds” practical password-hashing algorithm?

The answer is yes: compose a data-independent password-hashing algorithm (such as Balloon Hashing) with a data-dependent scheme (such as scrypt). To use the composed scheme, one would first run the password through the data-independent algorithm and next run the resulting hash through the data-dependent algorithm.\(^{12}\)

It is not difficult to show that the composed scheme is memory-hard against either: (a) an attacker who is able to learn the function’s data-access pattern on the target password, or (b) an attacker who mounts an attack in the pROM using the parallel algorithm of Alwen and Blocki [3]. The composed scheme defends against the two attacks separately but does not defend against both of them simultaneously: the composed function does not maintain memory-hardness in the face of an attacker who is powerful enough to get access-pattern information and mount a massively parallel attack. It would be far better to have a practical construction that could protect against both attacks simultaneously, but the best known algorithm that does this (the scheme of Alwen and Serbinenko [6]) is far too inefficient to use in practice.

The composed function is almost as fast as Balloon on its own—adding the data-dependent hashing function call is effectively as costly as increasing the round count of the Balloon algorithm by one.

5 Attacking and Defending Argon2

In this section, we analyze the Argon2i password hashing function [17], which won a recent password hashing competition [61].

An Attack. We first present an attack showing that it possible for an attacker to compute multi-pass Argon2i (the recommended version) saving a factor of

\(^{12}\) Our argument here gives some theoretical justification for the the Argon2id mode of operation proposed in some versions of the Argon2 specification [19, Appendix B]. That variant follows a hashing with a password-independent access pattern by hashing with a password-dependent access pattern.
\( e \approx 2.72 \) in space with no increase in computation time. Additionally, we show that an attacker can compute the single-pass variant of Argon2i, which is also described in the specification, saving more than a factor of four in space again with no increase in computation time. These attacks demonstrate unexpected weaknesses in the Argon2i design, and show the value of formal security analysis.

**A Defense.** In Appendix F we give the first proof of security showing that, with high probability, single-pass \( n \)-block Argon2i requires space \( S \) and time \( T \) to compute, such that \( S \cdot T \geq n^2/192 \). Our proof is relatively simple and uses the same techniques we have developed to reason about the Balloon algorithm. The time-space lower bound we can prove about Argon2i is weaker than the one we can prove about Balloon, since the Argon2i result leaves open the possibility of an attack that saves a factor of 192 factor in space with no increase in computation time. If Argon2i becomes standardized, it would be a worthwhile exercise to try to improve the constants on both the attacks and lower bounds to get a clearer picture of its exact memory-hardness properties.

### 5.1 Attack Overview

Our Argon2i attacks require a linear-time pre-computation operation that is independent of the password and salt. The attacker need only run the pre-computation phase once for a given choice of the Argon2i public parameters (buffer size, round count, etc.). After running the pre-computation step once, it is possible to compute many Argon2i password hashes, on different salts and different passwords using our small-space computation strategy. Thus, the cost of the pre-computation is amortized over many subsequent hash computations.

The attacks we demonstrate undermine the security claims of the Argon2i design documents. The design documents claim that computing \( n \)-block single-pass Argon2i with \( n/4 \) space incurs a 7.3× computational penalty [17, Table 2]. Our attacks show that there is no computational penalty. The design documents claim that computing \( n \)-block three-pass Argon2i with \( n/3 \) space incurs a 16,384× computational penalty [17, Section 5.4]. We compute the function in \( n/2.7 \approx n/3 \) space with no computational penalty.

We analyze a idealized version the Argon2i algorithm, which is slightly simpler than that proposed in the Argon2i v1.2.1 specification [17]. Our idealized analysis underestimates the efficacy of our small-space computation strategy, so the strategy we propose is actually more effective at computing Argon2i than the analysis suggests. The idealized analysis yields an expected \( n/4 \) storage cost, but as Figure 4 demonstrates, empirically our strategy allows computing single-pass Argon2i with only \( n/5 \) blocks of storage. This analysis focuses on the single-threaded instantiation of Argon2i—we have not tried to extend it to the many-threaded variant.

---

13 We have notified the Argon2i designers of this attack and the latest version of the specification incorporates a design change that attempts to prevent the attack [19]. We describe the attack on the original Argon2i design—we have not yet analyzed the revised algorithm.
5.2 Background on Argon.

At a high level, the Argon2i hashing scheme operates by filling up an \(n\)-block buffer with pseudo-random bytes, one 1024-byte block at a time. The first two blocks are derived from the password and salt. For \(i \in \{3, \ldots, n\}\), the block at index \(i\) is derived from two blocks: the block at index \((i - 1)\) and a block selected pseudo-randomly from the set of blocks generated so far. If we denote the contents of block \(i\) as \(x_i\), then Argon2i operates as follows:

\[
\begin{align*}
    x_1 &= H(\text{passwd}, \text{salt} \parallel 1) \\
    x_2 &= H(\text{passwd}, \text{salt} \parallel 2) \\
    x_i &= H(x_{i-1}, x_r) \quad \text{where } r_i \in \{1, \ldots, i - 1\}
\end{align*}
\]

Here, \(H\) is a non-memory-hard cryptographic hash function mapping two blocks into one block. The random index \(r_i\) is sampled from a non-uniform distribution over \(S_i = \{1, \ldots, i - 1\}\) that has a heavy bias towards blocks with larger indices. We model the index value \(r_i\) as if it were sampled from the uniform distribution over \(S_i\). Our small-space computation strategy performs better under a distribution biased towards larger indices, so our analysis is actually somewhat conservative.

The single-pass variant of Argon2i computes \((x_1, \ldots, x_n)\) in sequence and outputs bytes derived from the last block \(x_n\). Computing the function in the straightforward way requires storing every generated block for the duration of the computation—\(n\) blocks total.

The multiple-pass variant of Argon2i works as above except that it computes \(pn\) blocks instead of just \(n\) blocks, where \(p\) is a user-specified integer indicating the number of “passes” over the memory the algorithm takes. (The number of passes in Argon2i is analogous to number of rounds \(r\) in Balloon hashing.) The default number of passes is three. In multiple-pass Argon2i, the contents of block \(i\) are derived from the prior block and one of the most recent \(n\) blocks. The output of the function is derived from the value \(x_{pn}\). When computing the multiple-pass variant of Argon2i, one need only store the latest \(n\) blocks computed (since earlier blocks will never be referenced again), so the storage cost of the straightforward algorithm is still roughly \(n\) blocks.

Our analysis splits the Argon2i computation into discrete time steps, where time step \(t\) begins at the moment at which the algorithm invokes the compression function \(H\) for the \(t\)th time.

5.3 Attack Algorithm

Our strategy for computing \(p\)-pass Argon2i with fewer than \(n\) blocks of memory is as follows:

- **Pre-computation Phase.** We run the entire hash computation once—on an arbitrary password and salt—and write the memory access pattern to disk. For each memory block \(i\), we pre-compute the time \(t_i\) after which block
\( i \) is never again accessed and we store \( \{t_1, \ldots, t_{pn}\} \) in a read-only array. The total size of this table on a 64-bit machine is at most \( 8pn \) bytes.\(^{14}\) Since the Argon2i memory-access pattern does not depend on the password or salt, it is possible to use this same pre-computed table for many subsequent Argon2i hash computations (on different salts and passwords).

- **Computation Phase.** We compute the hash function as usual, except that we delete blocks that will never be accessed again. After reading block \( i \) during the hash computation at time step \( t \), we check whether the current time \( t \geq t_i \). If so, we delete block \( i \) from memory and reuse the space for a new block.

The expected space required to compute \( n \)-block single-pass Argon2i is \( n/4 \). The expected space required to compute \( n \)-block many-pass Argon2i tends to \( n/e \approx 2.7 \) as the number of passes tends to infinity. We analyze the space usage of the attack algorithm in detail in Appendix A.

### 6 Experimental Evaluation

In this section, we argue that the Balloon hashing algorithm is not only provably memory-hard; it is also practical. We first discuss how to compare the security properties of prior algorithms against those of Balloon. We then compare the performance of two existing algorithms (Argon2i and Catena) to the performance of Balloon.

\(^{14}\) On an FPGA or ASIC, this table can be stored in relatively cheap shared read-only memory and the storage cost can be amortized over a number of compute cores. Even on a general-purpose CPU, the table and memory buffer for the single-pass construction together will only require \( 8n + 1024(n/4) = 8n + 256n \) bytes when using our small-space computation strategy. Argon2i normally requires \( 1024n \) bytes of buffer space, so our strategy still yields a significant space savings.
6.1 How to Compare Memory-Hard Functions

The Catena password-hashing algorithm consists of two variants: BRG and DBG. It is possible to use a pebbling argument to prove the security of both Catena constructions, in the random-oracle model. The $r$-round $n$-block Catena DBG function has a claimed time-space lower bound of the form $S \cdot T \geq n \frac{2^n}{\text{poly}(n)}$, when $S \leq n/20$. The $n$-block Catena BRG constructions has a time-space lower bound of the form $S \cdot T \geq n^2$ (Catena BRG has no round parameter). As described in Section 3.2, the $n$-block $r$-round Balloon function has a time-space trade-off of the form $S \cdot T \geq (2^r - 1) \frac{n^2}{8}$ for $S < n/64$ when the parameter $\delta = 3$. How can we compare these functions, which exhibit such different time-space trade-offs?

One idea, which we adopt from the work of Alwen and Blocki [3], is to define the time-space advantage of an attacker at computing a function $f$. If we think of the time-space product as a proxy for the cost of a computation, then we can compare the attacker's time-space product to the defender's time-space product to learn whether the attacker is actually saving anything—in terms of computational cost—by mounting the attack.

If $A_S$ is an attacker using space $S$, we can let $ST_f(A_S)$ denote a lower-bound on the time-space product of $A_S$ at computing function $f$. We can let $ST_f(\text{Honest})$ denote an upper-bound on the time-space product of the “honest” algorithm for computing $f$ (i.e., the algorithm in the specification of $f$). We can then define the time-space advantage of $A_S$ at computing $f$ as

$$\text{Adv}[A_S, f] = \frac{ST_f(\text{Honest})}{ST_f(A_S)}.$$ 

If $\text{Adv}[A_S, f] \leq 1$, then we have $ST_f(\text{Honest}) \leq ST_f(A_S)$. In this case, the time-space product of the attacker exceeds that of the defender, so the attacker saves nothing over using the normal/honest algorithm for computing $f$. If $\text{Adv}[A_S, f] > 1$, then the attacker might be able to save some cost by running the attack.

Using this definition, we can bound the advantage of small-space attackers at computing Catena BRG and one-round Balloon:

$$\text{Adv}[A_S, \text{Catena-BRG}] \leq \frac{2n^2}{n^2/16} \leq 32; \quad \text{Adv}[A_S, \text{Balloon}_{(r=1)}] \leq \frac{2n^2}{n^2/8} \leq 16.$$ 

when $S \geq 128$. So the Balloon function offers a slightly stronger protection against small-space adversaries than does Catena BRG, but the two constructions are comparable.

A key difference between Catena BRG and Balloon hashing is that, by increasing the number of rounds of Balloon hashing, we can decrease a small-space attacker’s advantage to an arbitrarily small constant, whereas the attacker’s advantage for Catena BRG is fixed at 32.

$^{15}$ A pedantic technical point: the cost of computing the Balloon algorithm is actually $\delta rn^2$, but the factor of $\delta$ appears in the denominator here as well so we ignore both.
The comparison with Catena DBG is more interesting. Say that we want to set the parameters of our scheme to ensure that the advantage of an adversary using \( n/128 \) space is \( 1/8 \). That is, we want to penalize small-space adversaries with an \( 8 \times \) penalty in the space-time product required for the computation. For Balloon hashing, this requires:

\[
\text{Adv}[A_S, \text{Balloon}_r] \leq \frac{2n^2}{(2^r - 1)n^2}/8 \leq \frac{1}{8},
\]

so we can take \( r = 8 \) to bound the advantage in this way.

The honest strategy for computing Catena DBG requires \( 2n \) space for \( 2n \log_2 n \) time, so the advantage is

\[
\text{Adv}[A_S, \text{Catena-DBG}_r] \leq \frac{4n^2 \log_2 n}{n(\frac{n}{64})^r} \leq \frac{1}{8}.
\]

Since \( S = n/128 \), we need that \( 32n \log_2 n < (2r)^r \). If \( n = 2^{16} \), as might be the case for password hashing, we then need \( r \geq 7 \) for this advantage to hold. So, for these settings of the parameters, we can compare Balloon hashing with \( r = 8 \) to Catena DBG with \( r = 7 \) for comparing their performance. We use these settings in our comparison of the two functions (Figure 5).

As an aside: another interesting feature of the Balloon algorithm is that we can make a small-space attacker’s advantage asymptotically negligible by letting the number of rounds \( r \) grow with \( n \). If \( r = \log_2 n \), for example, then for \( S \leq n/8 \) and \( \delta = 7 \):

\[
\text{Adv}[A_S, \text{Balloon}_{\log_2 n}] \leq \frac{n^2 \log_2 n}{(n \log n - 1)n^2}/8 = \frac{8 \log_2 n}{(n \log n - 1)} \leq \text{negl}(n),
\]

for some function \( \text{negl}(\cdot) \) whose inverse grows faster than any fixed polynomial.

### 6.2 Experimental Set-up

We used the OpenSSL implementations (version 1.0.1f) of SHA-512 and the reference implementations of the other cryptographic hash functions (Blake2b, ECHO, and SHA-3/Keccak). We compiled the code for our timing results with gcc version 4.8.5 using the -O3 option. We use optimized versions of the underlying cryptographic primitives where available, but the core Balloon hash code is written entirely in C. Our source code is available at https://crypto.stanford.edu/balloon/ under the ISC open-source license. We used a workstation running an Intel Core i7-6700 CPU (Skylake) at 3.40 GHz with 8 GiB of RAM for our performance benchmarks. We average all of our measurements over 32 trials.

We compare the Balloon functions against Argon2i, which won the recent Password Hashing Competition (PHC) [61]. We also compare against the Catena hash function (DBG and BRG), which appeared at ASIACRYPT 2014 and was a PHC finalist [39]. Although Argon2i comes equipped with no formal security
analysis, we compare against it for performance reasons. For comparison purposes, we implemented the Argon2i, Catena BRG, and Catena DBG memory-hard algorithms in C. When we compare these algorithms against each other, we instantiate them all with Blake2b as the underlying cryptographic hash function [8].

6.3 Authentication Throughput

The goal of a memory-hard password hash function is to use as much working space as possible quickly as possible over the course of its computation. To evaluate the effectiveness of the Balloon algorithm on this metric, we measured the rate at which a server can check passwords (in hashes per second) for various buffer sizes on a single core.

Figure 5 shows the minimum buffer size required to compute each memory-hard function with high probability with no computational slowdown, for a variety of password hashing functions. We set the block size of the construction to be equal to the block size of the underlying compression function, to avoid the issues discussed in Appendix B.3. The charted results for Argon2i incorporate the fact that an adversary can compute many-pass Argon2i in a factor of \( e \approx 2.72 \) less working space than the defender must allocate for the computation and can compute single-pass Argon2i with a factor of four less space (see Section 5). For comparison, we also plot the space usage of two non-memory-hard password hashing functions, bcrypt [72] (with cost \( c = 12 \)) and PBKDF2-SHA512 [46] (with \( 10^5 \) iterations).

If we assume that an authentication server must perform 100 hashes per second per four-core machine, Figure 5 shows that it would be possible to use one-round Balloon hashing with a 2 MiB buffer or eight-round Balloon hashing with a 256 KiB buffer. At the same authentication rate, Argon2i (instantiated with Blake2b as the underlying cryptographic hash function) requires the attacker to use a smaller buffer—roughly 1.5 MiB for the one-pass variant. Thus, with Balloon hashing we simultaneously get better performance than Argon2i plus rigorous security guarantees.

6.4 Compression Function

Finally, Figure 6 shows the result of instantiating the Balloon algorithm construction with four different standard cryptographic hash functions: SHA-3 [16], Blake2b [8], SHA-512, and ECHO (a SHA-3 candidate that exploits the AES-NI instructions) [13]. The SHA-3 function (with rate = 1344) operates on 1344-bit blocks, and we configure the other hash functions to use 512-bit blocks.

On the x-axis, we plot the buffer size used in the Balloon function and on the y-axis, we plot the rate at which the Balloon function fills memory, in bytes of

---

\[ \] Argon2i uses a new (and very fast) hash function as its cryptographic compression function. Since Argon’s hash function is not even collision resistant, modeling it as a random oracle seems problematic, so we use a standard cryptographic hash function for our evaluation.
Fig. 5: The Balloon algorithm outperforms Argon2i and Catena DBG for many settings of the security parameters, and Balloon is competitive with Catena BRG. We instantiate Argon, Balloon, and Catena with Blake2b as the underlying cryptographic hash function.

Fig. 6: Throughput for the Balloon algorithm when instantiated with different compression functions. The dotted lines indicate the sizes of the L1, L2, and L3 caches on our test machine.
written per second. As Figure 6 demonstrates, Blake2b and ECHO outperform the SHA functions by a bit less than a factor of two.

7 Related Work

Password Hashing. The problem of how to securely store passwords on shared computer systems is nearly as old as the systems themselves. In a 1974 article, Evans et al. described the principle of storing passwords under a hard-to-invert function [37]. A few years later, Robert Morris and Ken Thompson presented the now-standard notion of password salts and explained how to store passwords under a moderately hard-to-compute one-way function to increase the cost of dictionary attacks [56]. Their DES-based “crypt” design became the standard for password storage for over a decade [52] and even has a formal analysis by Wagner and Goldberg [87].

In 1989, Feldmeier and Karn found that hardware improvements had driven the cost of brute-force password guessing attacks against DES crypt down by five orders of magnitude since 1979 [38, 49]. Poul-Henning Kamp introduced the costlier md5crypt to replace crypt, but hardware improvements also rendered that design outmoded [28].

Provos and Mazières realized that, in the face of ever-increasing processor speeds, any fixed password hashing algorithm would eventually become easy to compute and thus ineffective protection against dictionary attacks. Their solution, bcrypt, is a password hashing scheme with a variable “hardness” parameter [72]. By periodically ratcheting up the hardness, a system administrator can keep the time needed to compute a single hash roughly constant, even as hardware improves. A remaining weakness of bcrypt is that it exercises only a small fraction of the CPU’s resources—it barely touches the L2 and L3 caches during its execution [53]. To increase the cost of custom password-cracking hardware, Reinhold’s HEKS hash [73] and Percival’s popular scrypt routine consume an adjustable amount of storage space [65], in addition to time, as they compute a hash. Balloon, like scrypt, aims to be hard to compute in little space. Unlike scrypt, however, we require that our functions’ data access pattern be independent of the password to avoid leaking information via cache-timing attacks [22, 59, 84] (see also the attack in Section 4.2). The Dogecoin and Litecoin [21] crypto-currencies have incorporated scrypt as an ASIC-resistant proof-of-work function.

The recent Password Hashing Competition motivated the search for memory-hard password-hashing functions that use data-independent memory access patterns [61]. The Argon2 family of functions, which have excellent performance and an appealingly simple design, won the competition [17]. The Argon2 functions lack a theoretical analysis of the feasible time-space trade-offs against them; using the same ideas we have used to analyze the Balloon function, we provide the first such result in Appendix F.

The Catenahash functions [39], which became finalists in the Password Hashing Competition, are memory-hard functions whose analysis applies pebbling ar-
guments to classic graph-theoretic results of Lengauer and Tarjan [51]. The Balloon analysis we provide gives a tighter time-space lower bounds than Catena’s analysis can provide in many cases, and the Balloon algorithm outperforms the more robust of the two Catena algorithms (see Section 6).

The other competition finalists included a number of interesting designs that differ from ours in important ways. Makwa [70] supports offloading the work of password hashing to an untrusted server but is not memory-hard. Lyra [2] is a memory-hard function but lacks proven space-time lower bounds. Yescrypt [66] is an extension of scrypt and uses a password-dependent data access pattern.

Ren and Devadas [74] give an analysis of the Balloon algorithm using bipartite expanders, following the pebbling techniques of Paul and Tarjan [63]. Their results imply that an adversary that computes the $n$-block $r$-round Balloon function in $n/8$ space, must use at least $2^r n/c$ time to compute the function (for some constant $c$), with high probability in the random-oracle model. We prove the stronger statement that an adversary’s space-time product must satisfy: $S \cdot T \in \Omega(n^2)$ for almost all values of $S$. Ren and Devadas also prove statements showing that algorithms computing the Balloon functions efficiently must use a certain amount of space at many points during their computation. Our time-space lower bounds only show that the adversary must use a certain amount of space a some point during the Balloon computation.

Other Studies of Password Protection. Concurrently with the design of hashing schemes, there has been theoretical work from Bellare et al. on new security definitions for password-based cryptography [11] and from Di Crescenzo et al. on an analysis of passwords storage systems secure against adversaries that can steal only a bounded number of bits of the password file [30]. Other ideas for modifying password hashes include the key stretching schemes of Kelsey et al. [47] (variants on iterated hashes), a proposal by Boyen to keep the hash iteration count (e.g., time parameter in bcrypt) secret [23], a technique of Canetti et al. for using CAPTCHAs in concert with hashes [25], and a proposal by Dürmuth to use password hashing to do meaningful computation [31].

Parallel Memory-Hardness. In a recent line of work, Alwen and Serbinenko [6], Alwen et al. [4], and Alwen and Blocki [3] have analyzed memory-hard functions from a number of angles in the parallel random-oracle model. We discuss these very relevant results at length in Section 4.

Memory-Bound Functions. Abadi et al. [1] introduced memory-bound functions as more effective alternatives to traditional proofs-of-work in heterogeneous computing environments [9, 33]. These functions require many cache misses to compute and, under the assumption that memory latencies are consistent across computing platforms, they are roughly as hard to compute on a computationally powerful device as on a computationally weak one. The theoretical analysis of memory-bound functions represented one of the first applications of pebbling arguments to cryptography [32, 34].

Proofs of Space. Dziembowski et al. [35] and Ateniese et al. [7] study proofs-of-space. In these protocols, the prover and verifier agree on a large bitstring that
the prover is supposed to store. Later on, the prover can convince the verifier that the prover has stored the string on disk, even if the verifier does not store the string herself. Spacemint proposes building a cryptocurrency based upon a proof-of-space rather than a proof-of-work [60]. Ren and Devadas propose using the problem of pebbling a Balloon graph as the basis for a proof of space [74].

Four features distinguish a proof-of-space from a memory-hard password hashing function. First, a proof-of-space may access memory in a pattern that depends on the input to the proof. In the context of password hashing, we prefer to avoid data-dependent addressing to prevent cache attacks. Second, proofs-of-space must be efficiently checkable—the verifier of the proof should be quite efficient while the prover need not be. We need no such asymmetry in our setting: computing a hash should be moderately hard for all parties. Third, the output of a proof-of-space need not be pseudorandom, while password hashing functions (when used for key derivation) should have pseudorandom output. Fourth, a password hashing function should hide the entire input, whereas a proof-of-space function need not.

Time-Space Trade-Offs. The techniques we use to analyze Balloon draws on extensive prior work on computational time-space trade-offs. We use pebbling arguments, which have seen application to register allocation problems [78], to the analysis of the relationships between complexity classes [14, 26, 45, 82], and to prior cryptographic constructions [34, 35, 36, 39]. Pebbling has also been a topic of study in its own right [51, 64]. Savage’s text gives a clear introduction to graph pebbling [77] and Nordström surveys the vast body of pebbling results in depth [58].

8 Conclusion

We have introduced the Balloon password hashing algorithm. The Balloon algorithm is provably memory-hard (in the single-instance setting in the random-oracle model), exhibits a password-independent memory access pattern, and meets or exceeds the performance of the fastest heuristically secure schemes. Using a novel combinatorial pebbling argument, we have demonstrated that password-hashing algorithms can have rigorous memory-hardness proofs without sacrificing practicality.

This work raises a number of open questions:

- Are there efficient methods to defend against cache attacks on scrypt (Section 4.2)? Could a special-purpose ORAM scheme help [41]?
- Are there practical memory-hard functions with password-independent access patterns that retain their memory-hardness properties under parallel attacks [6]? Alwen and Blocki [3] show that an “optimally” memory-hard function cannot exist, but a near-optimal construction that is usable in practice would still be interesting.
- Are there simple techniques for arguing about the security of memory-hard functions (with data-independent access patterns) under parallel attacks?
Acknowledgements  We would like to thank Josh Benaloh, Joe Bonneau, Greg Hill, Ali Mashitizadeh, David Mazières, Yan Michalevsky, Bryan Parno, Greg Valiant, Riad Wahby, Keith Winstein, David Wu, Sergey Yekhanin, and Greg Zaverucha for comments on early versions of this work.

This work was funded in part by an NDSEG Fellowship, NSF, DARPA, a grant from ONR, and the Simons Foundation. Opinions, findings and conclusions or recommendations expressed in this material are those of the authors and do not necessarily reflect the views of DARPA.

Bibliography

[61] Password hashing competition. https://password-hashing.net/
A Details of the Attack on Argon2

In this section, we provide a detailed analysis of the attack on Argon2i that we introduced in Section 5.

The goal of the attack algorithm is to compute Argon2i in the same number of time steps as the naïve algorithm uses to compute the function, while using a constant factor less space than the naïve algorithm does. In this way, an attacker mounting a dictionary attack against a list of passwords hashed with Argon2i can do so at less cost (in terms of the space-time product) than the Argon2i specification claimed possible.

Argon2i has one-pass and many-pass variants and our attack applies to both; the many-pass variant is recommended in the specification. We first analyze the attack on the one-pass variant and then analyze the attack on the many-pass variant.

We are interested in the attack algorithm’s expected space usage at time step $t$—call this function $S(t)$.

Analysis of One-Pass Argon2i. At each step of the one-pass Argon2i algorithm, the expected space usage $S(t)$ is equal to the number of memory blocks generated so far minus the expected number of blocks in memory that will never be used after time $t$. Let $A_{i,t}$ be the event that block $i$ is never needed after time step $t$ in the computation. Then $S(t) = t - \sum_{i=1}^{t} \Pr[A_{i,t}]$.

To find $S(t)$ explicitly, we need to compute the probability that block $i$ is never used after time $t$. We know that the probability that block $i$ is never used after time $t$ is equal to the probability that block $i$ is not used at time $t+1$ and is not used at time $t+2$ and $\ldots$ and is not used at time $n$. Let $U_{i,t}$ denote the event that block $i$ is unused at time $t$. Then:

$$\Pr[A_{i,t}] = \Pr\left[\bigcap_{t'=t+1}^{n} U_{i,t'}\right] = \prod_{t'=t+1}^{n} \Pr[U_{i,t'}]$$  \hspace{1cm} (1)

The equality on the right-hand side comes from the fact that $U_{i,t'}$ and $U_{i,t''}$ are independent events for $t' \neq t''$.

To compute the probability that block $i$ is not used at time $t'$, consider that there are $t'-1$ blocks to choose from and $t'-2$ of them are not block $i$: $\Pr[U_{i,t'}] = \frac{t'-2}{t'-1}$. Plugging this back into Equation 1, we get:

$$\Pr[A_{i,t}] = \prod_{t'=t+1}^{n} \left(\frac{t'-2}{t'-1}\right) = \frac{t-1}{n-1}$$

17 As described in Section 5.2, the contents of block $i$ in Argon2i are derived from the contents of block $i-1$ and a block chosen at random from the set $r_i \in \{1, \ldots, i-1\}$. Throughout our analysis, all probabilities are taken over the random choices of the $r_i$ values.
Now we substitute this back into our original expression for $S(t)$:

$$S(t) = t - \sum_{i=1}^{t} \left( \frac{t - 1}{n - 1} \right) = t - \frac{t(t - 1)}{n - 1}$$

Taking the derivative $S'(t)$ and setting it to zero allows us to compute the value $t$ for which the expected storage is maximized. The maximum is at $t = n/2$ and the expected number of blocks required is $S(n/2) \approx n/4$.

**Larger in-degree.** A straightforward extension of this analysis handles the case in which $\delta$ random blocks—instead of one—are hashed together with the prior block at each step of the algorithm. Our analysis demonstrates that, even with this strategy, single-pass Argon2i is vulnerable to pre-computation attacks. The maximum space usage comes at $t^* = n/(\delta + 1)^{1/\delta}$, and the expected space usage over time $S(t)$ is:

$$S(t) \approx t - \frac{t^{\delta+1}}{n^\delta} \quad \text{so} \quad S(t^*) \approx \frac{\delta}{(\delta + 1)^{1+1/\delta}} n.$$ 

**Analysis of Many-Pass Argon2i.** One idea for increasing the minimum memory consumption of Argon2i is to increase the number of passes that the algorithm takes over the memory. For example, the Argon2 specification proposes taking three passes over the memory to protect against certain time-space tradeoffs. Unfortunately, even after many passes over the memory, the Argon2i algorithm sketched above still uses many fewer than $n$ blocks of memory, in expectation, at each time step.

To investigate the space usage of the many-pass Argon2i algorithm, first consider that the space usage will be maximized at some point in the middle of its computation—not in the first or last passes. At some time step $t$ in the middle of its computation the algorithm will have at most $n$ memory blocks in storage, but the algorithm can delete any of these $n$ blocks that it will never need after time $t$.

At each time step, the algorithm adds a new block to the end of the buffer and deletes the first block. At any one point in the algorithm’s execution, there will be at most $n$ blocks of memory in storage. If we freeze the execution of the Argon2i algorithm in the middle of its execution, we can inspect the $n$ blocks it has stored in memory. Call the first block “stored block 1” and the last block “stored block $n$.”

Let $B_{i,t}$ denote the event that stored block $i$ is never needed after time $t$. Then we claim $\Pr[B_{i,t}] = (\frac{n-1}{n})^i$. To see the logic behind this calculation: notice that, at time $t$, the first stored block in the buffer can be accessed at time $t + 1$ but by time $t + 2$, the first stored block will have been deleted from the buffer. Similarly, the second stored block in the buffer at time $t$ can be accessed at time $t + 1$ or $t + 2$, but not at time $t + 3$ (since by then stored block 2 will have been deleted from the buffer). Similarly, stored block $i$ can be accessed at time steps $(t + 1)$, $(t + 2)$, $\ldots$, $(t + i)$ but not at time step $(t + i + 1)$. 

31
The total storage required is then:

\[ S(t) = n - \sum_{i=1}^{n} E[B_{i,t}] = n - \sum_{i=1}^{n} \left( \frac{n - 1}{n} \right)^i \approx n - n \left( 1 - \frac{1}{e} \right). \]

Thus, even after many passes over the memory, Argon2i can still be computed in roughly \( n/e \) space with no time penalty.

B Pebble Games

In this section, we introduce pebble games and explain how to use them to analyze the Balloon algorithm.

B.1 Rules of the Game

The pebble game is a one-player game that takes place on a directed acyclic graph \( G = (V, E) \). If there is an edge \((u, v) \in E\), we say that \( v \) is a successor of \( u \) in the directed graph and that \( u \) is a predecessor of \( v \). As usual, we refer to nodes of the graph with in-degree zero as source nodes and nodes with out-degree zero as sink nodes—edges point from sources to sinks.

A pebbling game, defined next, represents a computation. The pebbles represent intermediate values stored in memory.

**Definition 2 (Pebbling game).** A pebbling game on a directed acyclic graph \( G = (V, E) \) consists of a sequence of player moves, where each move is one of the following:

- place a pebble on a source vertex,
- remove a pebble from any pebbled vertex, or
- place a pebble on a non-source vertex if and only if all of its predecessor vertices are pebbled.

The game ends when a pebble is placed on a designated sink vertex. A sequence of pebbling moves is legal if each move in the sequence obeys the rules of the game.

The pebble game typically begins with no pebbles on the graph, but in our analysis we will occasionally define partial pebblings that begin in a particular configuration \( C \), in which some some vertices are already pebbled.

The pebble game is a useful model of oblivious computation, in which the data-access pattern is independent of the value being computed [69]. Edges in the graph correspond to data dependencies, while vertices correspond to intermediate values needed in the computation. Source nodes represent input values (which have no dependencies) and sink nodes represent output values. The pebbles on the graph correspond to values stored in the computer’s memory at a point in the computation. The three possible moves in the pebble game then correspond to: (1) loading an input value into memory, (2) deleting a value stored in memory, and (3) computing an intermediate value from the values of its dependencies.
B.2 Pebbling in the Random-Oracle Model

Dwork, Naor, and Wee [34] demonstrated that there is a close relationship between the pebbling problem on a graph \( G \) and the problem of computing a certain function \( f_G \) in the random-oracle model [12]. This observation became the basis for the design of the Catena memory-hard hash function family [39] and is useful for our analysis as well.

Since the relationship between \( G \) and \( f_G \) will be important for our construction and security proofs, we will summarize here the transformation of Dwork et al. [34], as modified by Alwen and Serbinenko [6]. The transformation from directed acyclic graph \( G = (V,E) \) to function \( f_G \) works by assigning a label to each vertex \( v \in V \), with the aid of a cryptographic hash function \( H \).

Definition 3 (Labeling). Let \( G = (V,E) \) be a directed graph with maximum in-degree \( \delta \) and a unique sink vertex, let \( x \in \{0,1\}^k \) be a string, and let \( H : \mathbb{Z}_{|V|} \times (\{0,1\}^k \cup \{\bot\})^\delta \to \{0,1\}^k \) be a function, modeled as a random oracle. We define the labeling of \( G \) relative to \( H \) and \( x \) as:

\[
\text{label}_x(v_i) = \begin{cases} 
H(i,x,\bot,\ldots,\bot) & \text{if } v_i \text{ is a source} \\
H(i,\text{label}_x(z_1),\ldots,\text{label}_x(z_\delta)) & \text{o.w.}
\end{cases}
\]

where \( z_1,\ldots,z_\delta \) are the predecessors of \( v_i \) in the graph \( G \). If \( v_i \) has fewer than \( \delta \) predecessors, a special “empty” label (\( \bot \)) is used as placeholder input to \( H \).

The labeling of the graph \( G \) proceeds from the sources to the unique sink node: first, the sources of \( G \) receive labels, then their successors receive labels, and so on until finally the unique sink node receives a label. To convert a graph \( G \) into a function \( f_G : \{0,1\}^k \to \{0,1\}^k \), we define \( f_G(x) \) as the function that outputs the label of the unique sink vertex under the labeling of \( G \) relative to a hash function \( H \) and an input \( x \).

Dwork et al. demonstrate that any valid pebbling of the graph \( G \) with \( S \) pebbles and \( T \) placements immediately yields a method for computing \( f_G \) with \( Sk \) bits of space and \( T \) queries to the random oracle. Thus, upper bounds on the pebbling cost of a graph \( G \) yield upper bounds on the computation cost of the function \( f_G \). In the other direction, they show that with high probability, an algorithm for computing \( f_G \) with space \( Sk \) and \( T \) random oracle queries yields a pebbling strategy for \( G \) using roughly \( S \) pebbles and \( T \) placements [34, Lemma 1].

This piece of the argument is subtle, since an adversarial algorithm for computing \( f_G \) could store parts of labels, might try to guess labels, or might use some other arbitrary strategy to compute the labeling. Showing that every algorithm that computing \( f_G \) with high probability yields a pebbling requires handling all of these possible cases.
We use a version of their result due to Dziembowski et al. [36]. The probabilities in the following theorems are over the choice of the random oracle and the randomness of the adversary.

**Theorem 4 (Adapted from Theorem 4.2 of Dziembowski et al. [36])** Let $G$, $H$, and $k$ be as in Definition 3. Let $A$ be an adversary making at most $T$ random-oracle queries during its computation of $f_G(x)$. Then, given the sequence of $A$’s random oracle queries, it is possible to construct a pebbling strategy for $G$ with the following properties:

1. The pebbling is legal with probability $1 - T/2^k$.
2. If $A$ uses at most $Sk$ bits of storage then, for any $\lambda > 0$, the number of pebbles used is at most $\frac{Sk+\lambda}{k - \log_2 T}$, with probability $1 - 2^{-\lambda}$.
3. The number of pebble placements (i.e., moves in the pebbling) is at most $T$.

**Proof.** The first two parts are a special case of Theorem 4.2 of Dziembowski et al. [36]. To apply their theorem, consider the running time and space usage of the algorithm $A_{small}$ they define, setting $c = 0$. The third part of the theorem follows immediately from the pebbling they construct in the proof: there is at most one pebble placement per oracle call. There are at most $T$ oracle calls, so the total number of placements is bounded by $T$.

Informally, the lemma states that an algorithm using $Sk$ bits of space will rarely be able to generate a sequence of random oracle queries whose corresponding pebbling places more than $S$ pebbles on the graph or makes an invalid pebbling move.

The essential point, as captured in the following theorem, is that if we can construct a graph $G$ that takes a lot of time to pebble when using few pebbles, then we can construct a function $f_G$ that requires a lot of time to compute with high probability when using small space, in the random oracle model.

**Theorem 5** Let $G$ and $k$ be as in Definition 3 with the additional restriction that there is no pebbling strategy for $G$ using $S^*$ pebbles and $T^*$ pebble placements, where $T^*$ is less than $2^k - 1$. Let $A$ be an algorithm that makes $T$ random oracle queries and uses $\sigma$ bits of storage space. If

$$T < T^* \quad \text{and} \quad \sigma < S^*(k - \log_2 T^*) - k,$$

then $A$ correctly computes $f_G(\cdot)$ with probability at most $\frac{T+1}{2^k}$.

**Proof.** Fix an algorithm $A$ as in the statement of the theorem. By Theorem 4, from a trace of $A$’s execution we can extract a pebbling of $G$ that:

- is legal with probability at least $1 - T/2^k$,
- uses at most $\frac{\sigma + k}{k - \log_2 T^*} < S^*$ pebbles with probability at least $1 - 2^{-k}$, and
- makes at most $T$ pebble placements.
By construction of $G$, there does not exist a pebbling of $G$ using $S^*$ pebbles and $T^*$ pebble placements. Thus, whenever $\mathcal{A}$ succeeds at computing $f_G(\cdot)$ it must be that either (1) the pebbling we extract from $\mathcal{A}$ is invalid, (2) the pebbling we extract from $\mathcal{A}$ uses more than $S^*$ pebbles, or (3) the pebbling we extract from $\mathcal{A}$ uses more than $T^*$ moves. From Theorem 4, the probability of the first event is at most $T/2^k$, the probability of the second event is at most $1/2^k$, and the probability of the third event is zero.

By the Union Bound, we find that:

$$\Pr[\mathcal{A} \text{ succeeds}] \leq \Pr[\text{pebbling is illegal}] + \Pr[\text{pebbling uses } \gt S^* \text{ pebbles}] + \Pr[\text{pebbling uses } \gt T^* \text{ time steps}].$$

Substituting in the probabilities of each of these events derived from Theorem 4, we find

$$\Pr[\mathcal{A} \text{ succeeds}] \leq \frac{T}{2^k} + \frac{1}{2^k} = \frac{T + 1}{2^k}.$$

\[\square\]

### B.3 Dangers of the Pebbling Paradigm

The beauty of the pebbling paradigm is that it allows us to reason about the memory-hardness of certain functions by simply reasoning about the properties of graphs. That said, applying the pebbling model requires some care. For example, it is common practice to model an infinite-domain hash function $H : \{0, 1\}^* \rightarrow \{0, 1\}^k$ as a random oracle and then to instantiate $H$ with a concrete hash function (e.g., SHA-256) in the actual construction.

When using a random oracle with an infinitely large domain in this way, the pebbling analysis can give misleading results. The reason is that Theorem 4 relies on the fact that when $H$ is a random oracle, computing the value $H(x_1, \ldots, x_n)$

![Fig. 7: A graph requiring $n + 1$ pebbles to pebble in the random-oracle model (left) requires $O(1)$ storage to compute when using a Merkle-Damgård hash function (right).](image-url)
requires that the entire string \((x_1, \ldots, x_n)\) be written onto the oracle tape (i.e., be in memory) at the moment when the machine queries the oracle.

In practice, the hash function \(H\) used to construct the labeling of the pebble graph is not a random oracle, but is often a Merkle-Damgård-style hash function \([29, 55]\) built from a two-to-one compression function \(C : \{0, 1\}^{2k} \rightarrow \{0, 1\}^k\) as
\[
H(x_1, \ldots, x_n) = C(x_n, C(x_{n-1}, \ldots, C(x_1, \perp) \ldots)).
\]
If \(H\) is one such hash function, then the computation of \(H(x_1, \ldots, x_n)\) requires at most a constant number of blocks of storage on the work and oracle tapes at any moment, since the Merkle-Damgård hash can be computed incrementally.

The bottom line is that pebbling lower bounds suggest that the labeling of certain graphs, like the one depicted in Figure 7, require \(\Theta(n)\) blocks of storage to compute with high probability in the random oracle model. However, when \(H\) is a real Merkle-Damgård hash function, these functions actually take \(\tilde{O}(1)\) space to compute. The use of incrementally computable compression functions has led to actual security weaknesses in candidate memory-hard functions in the past \([18, \text{Section 4.2}]\), so these theoretical weaknesses have bearing on practice.

This failure of the random oracle model is one of the very few instances in which a practical scheme that is proven secure in the random-oracle model becomes insecure after replacing the random oracle with a concrete hash function (other examples include \([10, 24, 42, 57]\)). While prior works study variants of the Merkle-Damgård construction that are indifferentiable from a random oracle \([27]\), they do not factor these space usage issues into their designs.

To sidestep this issue entirely, we use the random oracle only to model compression functions with a fixed finite domain (i.e., two-to-one compression functions) whose internal state size is as large as their output size. For example, we model the compression function of SHA-512 as a random oracle, but do not model the entire [infinite-domain] SHA-512 function as a random oracle. When we use a sponge function, like SHA-3 \([16]\), we use it as a two-to-one compression function, in which we extract only as many bits of output as the capacity of the sponge.

C A Combinatorial Lemma on “Well-Spread Sets”

In this section, we prove a combinatorial lemma that we will need for the analysis of the Balloon functions.

Let \(X = \{x_1, x_2, \ldots, x_m\}\) be a multiset of integers written in non-decreasing order, such that \(1 \leq x_i \leq n\) for all \(x_i \in X\). The elements of \(X\) partition the integers from 0 to \(n\) into a collection of segments
\[
(0, x_1], (x_1, x_2], (x_2, x_3], \ldots, (x_{m-1}, x_m].
\]

If we remove some subset of the segments, we can ask: what is the total length of the remaining segments? More formally, fix some set \(S \subseteq X\). Then, define the spread of the multiset \(X\) after removing \(S\) as the sum of the lengths of the
segments defined by $X$ whose rightmost endpoints are not in $S$. Symbolically, we can let $x_0 = 0$ and define the spread as:

$$\text{spread}_S(X) = \sum_{x_i \in X \setminus S} (x_i - x_{i-1}).$$

For example, we have that:

- $\text{spread}_{\emptyset}(\{2, 2, 4, 7\}) = (2 - 0) + (2 - 2) + (4 - 2) = 4$
- $\text{spread}_{\{7\}}(\{2, 2, 4, 7\}) = (2 - 0) + (2 - 2) = 2$
- $\text{spread}_{\{2, 7\}}(\{2, 2, 4, 7\}) = (4 - 2) = 2$
- $\text{spread}_{\{2, 4, 7\}}(\{2, 2, 4, 7\}) = 0$

In computing $\text{spread}_S(\cdot)$, we remove all segments whose right endpoint falls into the set $S$. If there are many such segments (i.e., as when we compute $\text{spread}_{\{7\}}(\{2, 2, 4, 7\}) = 5$), we remove all of them.

We say that a multiset of integers is a well-spread set if the spread of the set is larger than a certain threshold, even after removing any subset of segments of a particular size.

**Definition 6 (Well-Spread Set).** Let $X$ be a multiset of integers $X \subseteq \{1, \ldots, n\}$ and let $\sigma$ be an integer such that $\sigma \leq |X|$. Then $X$ is a $(\sigma, \rho)$-well-spread set if for all subsets $S \subseteq X$ of size $\sigma$, $\text{spread}_S(X) \geq \rho$.

So, for example, a set $X$ of integers is $(n/8, n/4)$-well-spread if, for all sets $S \subseteq X$ of size $n/8$, we have that $\text{spread}_S(X) \geq n/4$.

The following lemma demonstrates that a randomly sampled set of integers induces a well-spread set with all but negligible probability.

**Lemma 7 (Big Spread Lemma)** For all constants $\delta > 1$ and $\omega > 0$, positive integers $n$, and positive integers $m < (1 - 2^{-\omega})n$, a multiset of $\delta m$ elements sampled independently and uniformly at random from $\{1, \ldots, n\}$ is an $(m, n/2^{\omega})$-well-spread set with probability at least $1 - 2^{(1-\omega)\delta + \omega m}$.

Lemma 7 states (roughly) that if you throw $\delta m$ balls into $n$ bins, then remove the $m$ longest runs of empty bins, there are still at least $n/2^\omega$ bins left over, with very high probability. For example, if we take $\delta = 3$ and $\omega = 2$, the lemma states that if you throw $3m$ balls into $n$ bins (such that $m < 3n/4$) and then remove the $m$ longest runs of empty bins, there are still at least $n/4$ bins left over, with very high probability.

To prove Lemma 7, we need one related lemma. In the hypothesis of the following lemma, we divide the integers from 1 to $n$ into $k$ segments at random, with the $i$th segment having length $L_i$. The lemma states that if $f$ is an arbitrary function of the lengths of the segments, then the probability that $f$ takes on value 1 is invariant under a reordering of the segments.

**Lemma 8** Sample a set of $k$ integers independently and uniformly at random from $\{1, \ldots, n\}$. Write the elements of the set in non-decreasing order as
For all \( 1 \leq i \leq k \), define a random variable \( L_i = x_i - x_{i-1} \), with \( x_0 = 0 \). Then for all functions \( f : \mathbb{Z}^k \to \{0, 1\} \) and all permutations \( \pi \) on \( k \) elements,

\[
\Pr[f(L_1, \ldots, L_k) = 1] = \Pr[f(L_{\pi(1)}, \ldots, L_{\pi(k)}) = 1].
\]

To illustrate the meaning of the lemma, consider as an example the function \( f(L_3 > L_4) \) that takes the value “1” if the third segment is larger than the fourth segment and that takes the value “0” otherwise. The lemma implies that the probability that \( f(L_3 > L_4) \) takes on the value “1” is equal to the probability that the first segment is longer than the second segment.

**Proof of Lemma 8.** For convenience, write \( L = (L_1, \ldots, L_k) \). For a permutation \( \pi \) on \( k \) elements, let \( \pi(L) = (L_{\pi(1)}, \ldots, L_{\pi(k)}) \). The random variable \( L \) can take on value \( \ell \in \mathbb{Z}^k \) for any \( \ell \) such that \( \sum \ell_i \leq n \). We can rewrite the probability \( \Pr[f(L) = 1] \) by summing over the possible values \( \ell \in \mathbb{Z}^k \):

\[
\Pr[f(L) = 1] = \sum_{\ell} \Pr[f(L) = 1 | L = \ell] \cdot \Pr[L = \ell],
\]

\[
= \sum_{\ell} \Pr[f(\pi(L)) = 1 | \pi(L) = \ell] \cdot \Pr[L = \ell].
\]  

(2)

In the second step, we just renamed the variables on the right-hand side.

Now, we claim that \( \Pr[L = \ell] = \Pr[\pi(L) = \ell] \). The claim holds because there is a one-to-one correspondence between outcomes for which \( L = \ell \) and outcomes for which \( \pi(L) = \ell \). In particular, for each set of integers \( X = (x_1, x_2, \ldots, x_k) \) for which \( L = \ell \), there is a corresponding set \( X' = (x'_1, x'_2, \ldots, x'_k) \) for which \( \pi(L) = \ell \), and vice versa. We can write out the correspondence as follows:

\[
X = \{\ell_1, \ell_1 + \ell_2, \ell_1 + \ell_2 + \ell_3, \ldots\}
\]

\[
X' = \{\ell_{\pi^{-1}(1)}, \ell_{\pi^{-1}(1)} + \ell_{\pi^{-1}(2)}, \ell_{\pi^{-1}(1)} + \ell_{\pi^{-1}(2)} + \ell_{\pi^{-1}(3)}, \ldots\}.
\]

Thus, there are an equal number of outcomes for which \( L = \ell \) and \( \pi(L) = \ell \), and we conclude that \( \Pr[L = \ell] = \Pr[\pi(L) = \ell] \).

Finally, we substitute this last equation into (2) to get:

\[
\Pr[f(L) = 1] = \sum_{\ell} \Pr[f(\pi(L)) = 1 | \pi(L) = \ell] \cdot \Pr[\pi(L) = \ell]
\]

\[
= \Pr[f(\pi(L)) = 1].
\]

Having proved Lemma 8, we can return to prove Lemma 7.

**Proof of Lemma 7.** The set \( X \) partitions the integers from 1 to \( n \) into \( \delta m \) segments. (We ignore the rightmost segment, since we do not count it in the spread.) Let \( L_i \) be a random variable denoting the length of the \( i \)th segment.
The bad event $B$ that we want to avoid is the event that there exists some set of $m$ segments whose collective length is greater than $(1 - 2^{-\omega})n$. To bound $\Pr[B]$, we can bound the probability that there exists a subset $S \subseteq \{1, \ldots, \delta m \}$ of $m$ segments such that

$$\sum_{i \in S} L_i \geq (1 - 2^{-\omega})n.$$  

Fix a subset $S \subseteq \{1, \ldots, \delta m \}$ of size $m$. Let $f : \mathbb{Z}^{\delta m} \to \{0, 1\}$ be a function that outputs 1 if its first $m$ arguments sum to at least $(1 - 2^{-\omega})n$, and that outputs 0 otherwise. Let $\pi$ be a permutation on $\delta m$ elements such that if $i \in S$, then $L_i$ appears as one of the first $m$ elements of $(L_{\pi(1)}, \ldots, L_{\pi(\delta m)})$. Then,

$$\Pr \left[ \sum_{i \in S} L_i \geq (1 - 2^{-\omega})n \right] = \Pr [ f(L_{\pi(1)}, \ldots, L_{\pi(\delta m)}) = 1 ]$$

$$= \Pr [ f(L_1, \ldots, L_{\delta m}) = 1 ]$$

$$= \Pr [ L_1 + \cdots + L_m \geq (1 - 2^{-\omega})n ].$$  

We applied Lemma 8 to derive the second equality above.

The last probability on the right-hand side is relatively easy to compute. As long as $m < (1 - 2^{-\omega})n$, the event $L_1 + L_2 + \ldots + L_m \geq (1 - 2^{-\omega})n$ can only happen when $(\delta m - m)$ of the balls fall in the $2^{-\omega}n$ rightmost bins. (Otherwise the first $m$ segments would have length less than $(1 - 2^{-\omega})n$.) The probability of $(\delta m - m)$ balls falling in the rightmost $2^{-\omega}n$ bins is at most $(2^{-\omega})^{\delta m - m}$, so

$$\Pr[L_1 + L_2 + \ldots + L_m \geq (1 - 2^{-\omega})n] \leq \left( \frac{1}{2} \right)^{\omega(\delta m - m)}.$$  

Then we apply the Union Bound over all $\binom{\delta m}{m}$ possible size-$m$ subsets $S$ of segments that could be large to get:

$$\Pr[B] \leq \sum_{\text{sets } S} \Pr[L_1 + L_2 + \ldots + L_m \geq (1 - 2^{-\omega})n]$$

$$\leq \binom{\delta m}{m} \left( \frac{1}{2} \right)^{\omega(\delta - 1)m} \leq 2^{\delta m} \left( \frac{1}{2} \right)^{\omega(\delta - 1)m} \leq 2^{((1 - \omega)\delta + \omega)m}.$$

\[ \Box \]

\section{Sandwich Graphs}

In this section we recall the definition of sandwich graphs \cite{6}, and introduce a few transformations on sandwich graphs that are useful for our analysis of the Balloon functions.
D.1 Definitions

Definition 9 (Sandwich Graph [6]). A sandwich graph is a directed acyclic graph $G = (U \cup V, E)$ on $2n$ vertices, which we label as $U = \{u_1, \ldots, u_n\}$ and $V = \{v_1, \ldots, v_n\}$. The edges of $G$ are such that
- there is a $(u_i, u_{i+1})$ edge for $i = 1, \ldots, n - 1$,
- there is a $(u_n, v_1)$ edge,
- there is a $(v_i, v_{i+1})$ edge for $i = 1, \ldots, n - 1$, and
- all other edges cross from $U$ to $V$.

Figure 9 (left) depicts a sandwich graph. If $G = (U \cup V, E)$ is a sandwich graph, we refer to the vertices in $U$ as the “top” vertices and vertices in $V$ as the “bottom” vertices.

Definition 10 (Well-Spread Predecessors). Let $G = (U \cup V, E)$ be a sandwich graph and fix a subset of vertices $V' \subset V$. Write the immediate predecessors of $V'$ in $U$ as $P = \{u_{i_1}, u_{i_2}, \ldots, u_{i_{|P|}}\}$. Then we say that the predecessors of $V'$ are $(\sigma, \rho)$-well spread if the corresponding set of integers $\{i_1, i_2, \ldots, i_{|P|}\}$ is $(\sigma, \rho)$-well spread, in the sense of Definition 6.

Definition 11 (Avoiding Set). Let $G = (U \cup V, E)$ be a directed acyclic graph. We say that the subset $V' \subset V$ is a $(\sigma, \rho)$-avoiding set if, after placing $\sigma$ pebbles anywhere on the graph, except on $V'$, there are at least $\rho$ distinct vertices in $U$ on unpebbled paths to $V'$.

Fig. 8: The set $V'$ is a $(1, 4)$-avoiding set: after placing any single pebble on the graph (except on vertices in $V'$), there will still be at least four vertices in $U$ on unpebbled paths to $V'$.

Figure 8 gives an example of an avoiding set.

Lemma 12 Let $G = (U \cup V, E)$ be a sandwich graph and let $V \subset V'$. If the predecessors of $V'$ are $(\sigma, \rho)$-well spread, then $V'$ is a $(\sigma, \rho)$-avoiding set.

More formally: for all possible placements of $\sigma$ vertices on the graph $G$ except on $V'$, there exists a size-$\rho$ set of vertices $U' \subseteq U$ such that for all vertices $u \in U'$, there exists a vertex $v \in V'$ and a $u$-to-$v$ path $p$ in $G$ such that no vertex in the path $p$ contains a pebble.

40
Proof. We can think of the \( p \) predecessors of vertices in \( V' \) as dividing the chain of vertices in \( U \) (the “top half of the sandwich”) into \( p \) smaller sub-chains of vertices. If the predecessors of \( V' \) are \((\sigma, \rho)\)-well spread then, after removing any \( \sigma \) of these sub-chains, the remaining chains collectively contain \( \rho \) vertices. We know that there can be at most \( \sigma \) pebbles on vertices in \( U \), so at most \( \sigma \) sub-chains contain a pebbled vertex. The remaining sub-chains, collectively containing \( \rho \) vertices, contain no pebbled vertices. The \( \rho \) vertices in these unpebbled sub-chains all are on unpebbled paths to \( V' \), which proves the lemma.

Definition 13 (Everywhere-Avoiding Graph). Let \( G = (U \cup V, E) \) be a directed acyclic graph. The graph \( G \) is a \((\sigma, \rho)\)-everywhere avoiding graph if for every subset \( V' \subset V \) such that \(|V'| = \sigma\), the subset \( V' \) is a \((\sigma, \rho)\)-avoiding set.

Definition 14 (Consecutively-Avoiding Graph). Let \( G = (U \cup V, E) \) be a directed acyclic graph, with vertices in \( V \) labeled \( v_1 \) to \( v_n \). The graph \( G \) is a \((\kappa, \sigma, \rho)\)-consecutively avoiding graph if every subset \( V' \subset V \) of consecutive vertices of size \( \kappa \) is \((\sigma, \rho)\)-avoiding.

D.2 Transformations on Sandwich Graphs

Sandwich graphs are useful in part because they maintain certain connectivity properties under a “localizing” transformation. Let \( G \) be a sandwich graph. Let \( (a_1, \ldots, a_n) \) be the top vertices of the graph \( G \) and let \( (a_{n+1}, \ldots, a_{2n}) \) be the bottom vertices. The localized graph \( L(G) \) on vertices \((a_1, \ldots, a_{2n})\) has the property that every predecessor of a vertex \( a_i \) falls into the set the set \( \{a_{\max\{1,i-n\}}, \ldots, a_{i-1}\} \). (In the unlocalized graph, \( a_i \)’s predecessors fell into the larger set \( \{a_{\max\{1,i-2n\}}, a_{i-1}\}\).) If we think of the set \( \{a_{\max\{1,i-n\}}, \ldots, a_{i-1}\} \) as the vertices “nearby” to vertex \( a_i \), then the localizing transformation ensures that the predecessors of every vertex \( a_i \) all fall into this set of nearby vertices. Figure 9 demonstrates this transformation.

We use this localizing transformation to make more efficient use of buffer space in the Balloon algorithm. It is possible to pebble a localized sandwich graph in linear time with \( n + O(1) \) pebbles, whereas a non-localized sandwich graph can require as many as \( 2n \) pebbles in the worst case. This transformation makes computing the Balloon function easier for anyone using \( n \) space, while maintaining the property that the function is hard to compute in much less space. (Smith and Zhang find a similar locality property useful in the context of leakage-resilient cryptography [79].)

Definition 15. Let \( G = (U \cup V, E) \) be a sandwich graph with \( U = (u_1, \ldots, u_n) \), \( V = (v_1, \ldots, v_n) \). The localized graph \( L(G) = (L(V) \cup L(U), L(E)) \) has top and bottom vertex sets \( L(U) = \{\tilde{u}_1, \ldots, \tilde{u}_n\} \) and \( L(V) = \{\tilde{v}_1, \ldots, \tilde{v}_n\} \), and an edge
Claim 16 Let $G = (U \cup V, E)$ be a sandwich graph and let $V' \subset V$ be a subset whose predecessors are $(\sigma, \rho)$-well spread. Let $\mathcal{L}(V')$ be the vertices corresponding to $V'$ in the localized graph $\mathcal{L}(G)$. Then $\mathcal{L}(V')$ is a $(\sigma, \rho)$-avoiding set.

Proof. Fix a pebbling of $\mathcal{L}(G)$ using $\sigma$ pebbles with no pebbles on $V'$. For every edge $(u_i, v_j) \in U \times V$ in $G$, there is either (a) a corresponding edge in $\mathcal{L}(G)$, or (b) a pair of edges $(u_i, v_i)$ and $(v_i, v_j)$ in $\mathcal{L}(G)$. If the vertex $v_i$ does not contain a pebble, then for analyzing the avoiding-set property, we can consider there to exist a $(u_i, v_j)$ edge. There are now at most $\sigma$ pebbled $U$-to-$V'$ edges. By Lemma 12, the $(\sigma, \rho)$-avoiding set property follows.

Corollary 17 If $G$ is a sandwich graph such that every subset of $\sigma$ vertices $V' \subset V$ is $(\sigma, \rho)$-well-spread, then $\mathcal{L}(G)$ is a $(\sigma, \rho)$-everywhere avoiding graph.

Corollary 18 If $G$ is a sandwich graph such that every subset of $\kappa$ vertices $V' \subset V$ is $(\sigma, \rho)$-well-spread, then then $\mathcal{L}(G)$ is a $(\kappa, \sigma, \rho)$-consecutively avoiding graph.

The next set of graph transformations we use allows us to reduce the in-degree of the graph from $\delta$ down to 2 without affecting the key structural properties of the graph. Reducing the degree of the graph allows us to instantiate our construction with a standard two-to-one compression function and avoids the issues raised in Appendix B.3. The strategy we use follows the technique of Paul and Tarjan [63].
Definition 19. Let \( G = (U \cup V, E) \) be a (possibly localized) sandwich graph. We say that the degree-reduced graph \( D(G) \) is the graph in which each vertex \( v_i \in V \) in \( G \) of in-degree \( \delta + 1 \) is replaced with a path “gadget” whose vertices have in-degree at most 2. The original predecessor vertex \( v_i - 1 \) is at the beginning of the path, there are \( \delta - 1 \) internal vertices on the path, and the original vertex \( v_i \) is at the end of the path. The \( \delta \) other predecessors of \( v_i \) are connected to the vertices of the path (see Figure 10).

By construction, vertices in \( D(G) \) have in-degree at most two. If \( G \) is a sandwich graph on \( 2n \) vertices, then \( D(G) \) still has \( n \) “top” vertices and \( n \) “bottom” vertices. If the graph \( G \) had out-degree at most \( \delta \), then the vertex and edge sets of \( D(G) \) are at most a factor of \( (\delta - 1) \) larger than in \( G \), since each gadget has at most \( (\delta - 1) \) vertices. The degree-reduced graph \( D(G) \) has extra “middle” vertices (non-top non-bottom vertices) consisting of the internal vertices of the degree-reduction gadgets (Figure 10).

Claim 20  Let \( G = (U \cup V, E) \) be a (possibly localized) sandwich graph and let \( D(G) \) be the corresponding degree-reduced graph. Let \( V' \subset V \) be a \((\sigma, \rho)\)-avoiding set in \( G \), let \( D(V') \) be the vertices corresponding to \( V' \) and the degree-reduction gadgets attached to vertices in \( V' \). Then \( D(V') \) is a \((\sigma, \rho)\)-avoiding set.

Proof. For every pebbling of vertices in \( D(G) \) that violates the avoiding property of \( D(V') \), there is a corresponding pebbling in \( G \) that violates the avoiding property of \( V' \). To convert a pebbling of \( D(G) \) into a pebbling of \( G \), just place a pebble on every vertex in \( G \) whose corresponding degree-reduction gadget in \( D(G) \) has a pebble. The claim follows.

Corollary 21 If \( G \) is a \((\sigma, \rho)\)-everywhere avoiding sandwich graph, then \( D(G) \) is “almost” a \((\sigma, \rho)\)-everywhere avoiding graph, in the sense that every subset of \( \sigma \) degree-reduction gadgets is a \((\sigma, \rho)\)-avoiding set.

Corollary 22 If \( G \) is a \((\kappa, \sigma, \rho)\)-consecutively avoiding sandwich graph, then \( D(G) \) is “almost” \((\kappa, \sigma, \rho)\)-consecutively avoiding graph, in the sense that every subset of \( \sigma \) degree-reduction gadgets is a \((\sigma, \rho)\)-avoiding set.
D.3 Pebbling Sandwich Graphs

**Lemma 23** Let $G = (U \cup V, E)$ be a $(\kappa, \sigma, \rho)$-consecutively-avoiding sandwich graph on $2n$ vertices. Let $M$ be a legal sequence of pebbling moves that begins with no pebbles on the bottom half of the graph and that pebbles the topologically last vertex in $G$ at some point. Then we can divide $M$ into $L = n/(2\kappa)$ subsequences of legal pebbling moves $M_1, \ldots, M_L$ such that each subsequence $M_i$ pebbles at least $\rho$ unpebbled vertices in $U$.

The proof follows the idea of Lengauer and Tarjan’s analysis of pebbling strategies for the “bit-reversal” graph [51].

**Proof.** Label the vertices in $V$ in topological order as $(v_1, \ldots, v_n)$. Divide these vertices into $\lfloor n/\kappa \rfloor$ intervals of size $\kappa$. The last vertex in each of the intervals is then: $v_{\kappa}, v_{2\kappa}, v_{3\kappa}, \ldots, v_{\lfloor n/\kappa \rfloor \kappa}$.

Consider any legal sequence of pebbling moves $M$ of the last vertex in $G$. Let $t_0 = 0$ and let $t_i$ be the time step at which vertex $v_{i\kappa}$ (the last vertex in the $i$th interval) first receives a pebble. We know that at $t_{i-1}$, there are no pebbles on the $i$th interval—this is because, by the structure of a sandwich graph, all of the pebbles in $V$ must be pebbled in order. Thus, between $t_{i-1}$ and $t_i$, every dependency of the vertices in interval $i$ must receive a pebble. These are $\kappa$ consecutive vertices. Since $G$ is $(\kappa, \sigma, \rho)$-consecutively avoiding, the vertices in interval $i$ are a $(\sigma, \rho)$-avoiding set. By the definition of an avoiding set, at time $t_{i-1}$ there must be at least $\rho$ unpebbled dependencies in $U$ of the vertices in the $i$th interval. All of these vertices must receive pebbles by time $t_i$. Thus, in each time interval, the $M$ must pebble a set of $\rho$ unpebbled vertices in $U$.

We have that $\kappa \leq n$, so $\lfloor n/\kappa \rfloor \geq n/(2\kappa)$, so there are at least $n/(2\kappa)$ sets of $\rho$ unpebbled vertices in $U$ that receive pebbles during the pebbling.

We now let the subsequence $M_i$ defined in the lemma be the sequence of pebbling moves between $t_{i-1}$ and $t_i$, and the lemma follows.

**Corollary 24** Pebbling a $(\kappa, \sigma, \rho)$-consecutively avoiding (possibly localized) sandwich graph with $\sigma$ pebbles, starting with no pebbles on the bottom half of the graph requires at least $\frac{\sigma}{2\kappa}$ pebbling moves.

**Proof.** By Lemma 23, the pebbling pebbles at least $\rho$ unpebbled vertices at least $n/(2\kappa)$ times, so the total time required must be at least $\frac{\sigma}{2\kappa}$.

A key piece of our analysis involves “stacks” of everywhere-avoiding sandwich graphs. Given a sandwich graph with $n$ top vertices and $n$ bottom vertices, we can stack the graphs by making the bottom nodes of the zero-th copy of the graph, the top nodes of the first copy of the graph (see Figure 11).

**Lemma 25** Let $G$ be a depth-$d$ stack of $(\sigma, 2\sigma)$-everywhere avoiding sandwich graphs. Let $V'$ be a set of $\sigma$ level-$d$ vertices in the graph. Fix a configuration $C$ of at most $\sigma$ pebbles anywhere on the graph, except that there are no pebbles on vertices in $V'$. Then if $M$ is a sequence of legal pebbling moves such that...
Fig. 11: A stack of $d = 3$ sandwich graphs. The top vertices of the stack are at level $L_0$ and the bottom vertices are at level $L_3$.

- $M$ beings in configuration $C$,
- $M$ at some point places a pebble on every vertex in $V'$, and
- $M$ never uses more than $\sigma$ pebbles,

then $M$ must consist of at least $2^d \sigma$ pebbling moves.

**Proof.** By induction on the depth $d$.

**Base case** ($d = 1$). By the fact that the graph $G$ is $(\sigma, 2\sigma)$-everywhere avoiding, the $\sigma$ vertices in the set $V'$ must have at least $2\sigma$ unpebbled dependencies on the top level of the graph. These unpebbled predecessors of vertices in $V'$ all must receive pebbles during $M$, so $M$ must contain at least $2\sigma$ moves.

**Induction Step.** As in the base case, we know that there are at least $2\sigma$ unpebbled level-$(d - 1)$ dependencies of $V'$ that must receive pebbles during $M$. Now we can divide $M$ into two sub-sequences of consecutive pebbling moves: $M = M_1 \parallel M_2$. We divide the pebbling moves such that $M_1$ consists of the moves during which the first $\sigma$ of the $2\sigma$ unpebbled dependencies receive pebbles, and $M_2$ consists of the rest of the moves.

Note now that the induction hypothesis applies to both $M_1$ and $M_2$:

- Each set of moves begins in a configuration of at most $\sigma$ pebbles on the graph.
- Each set of moves pebbles a set of $\sigma$ initially unpebbled level-$(d - 1)$ vertices.
- Each set of moves never uses more than $\sigma$ pebbles.

Thus the total number of pebbling moves required in $M$ is at least $2 \cdot (2^{d - 1} \sigma)$, which proves the lemma.

**Remark 26** The arguments of Lemmata 23 and 25 apply also to localized and degree-reduced sandwich graphs, by Corollaries 17, 18, 21, and 22.
D.4 Random Sandwich Graphs

Now we introduce two special types of sandwich graphs that we use in the analysis of the Balloon function. Alwen and Blocki [3] use the same type of sandwich graphs in their analysis (indeed, their work has inspired our use of sandwich graphs), though the results we prove here are new.

**Definition 27 (δ-Random Sandwich Graph).** A δ-random sandwich graph is a sandwich graph $G = (U \cup V, E)$ such that each vertex in $V$ has δ predecessors sampled independently and uniformly at random from the set $U$.

**Lemma 28** Fix integers $\delta \geq 3$ and $\omega \geq 2$. Let $G = (U \cup V, E)$ be a δ-random sandwich graph on $2n$ vertices. Then for all positive integers $n_0$, $G$ is an $(m, m, n/2^\omega)$-consecutively avoiding graph for all $m$ such that $n_0 \leq m < (1 - 2^{-\omega})n$, except with probability $2n \cdot c_{28}(\delta, \omega)^{n_0}$, where the value of $c_{28}(\cdot, \cdot)$ depends only on $\delta$ and $\omega$.

**Proof.** The predecessors of a set of $m = |V'|$ vertices are $\delta m$ vertices i.u.r. sampled from the set of $n$ “top” vertices. The probability that a single consecutive set of $m$ vertices induces a poorly spread set of predecessors is at most $2(1 - 2^{-\omega})\delta + \omega$, by Lemma 7. We then apply the Union Bound over all sets of $m$ consecutive vertices to find the probability that there exists a bad set. For each choice of $m$, there are at most $n$ sets $V'$ of consecutive vertices, so

$$\Pr[\exists \text{ bad set}] \leq \sum_{i=1}^{n} \sum_{m=1}^{n} (2(1 - 2^{-\omega})\delta + \omega)^m \leq \sum_{m=n_0}^{\infty} (2(1 - 2^{-\omega})\delta + \omega)^m.$$  

This is just a geometric series with ratio $r = 2(1 - 2^{-\omega})\delta + \omega$.

$$\Pr[\exists \text{ bad set}] \leq n \cdot \frac{r^{n_0}}{1 - r} \leq n \cdot \frac{2(1 - 2^{-\omega})\delta + \omega |n_0|}{1 - 2(1 - 2^{-\omega})\delta + \omega}$$

Since $\delta \geq 3$ and $\omega \geq 2$, the denominator is at least $1/2$, so

$$\Pr[\exists \text{ bad set}] \leq 2n \cdot 2(1 - 2^{-\omega})\delta + \omega |n_0|.$$  

Taking $c_{28}(\delta, \omega) = 2(1 - 2^{-\omega})\delta + \omega$ proves the lemma. \qed

**Lemma 29** For every $\omega \geq 2$, there exists constants $\delta \geq 1$ and $c < 1$ such that a δ-random sandwich graph on $2n$ vertices is an $(n/2^{\omega+1}, n/2^{\omega})$-everywhere avoiding graph, with probability at least $1 - c_{29}(\delta, \omega)^n$, for some function $c_{29}(\cdot, \cdot)$.

**Proof.** The probability that a single set $V' \subset V$ in $G$ induces a set that is not $(n/2^{\omega+1}, n/2^{\omega})$-well spread is at most $2(1 - 2^{-\omega})\delta + \omega n/2^{\omega+1}$, by Lemma 7. We then apply the Union Bound over sets $V'$ of size $n/2^{\omega+1}$ to find the probability that
there exists a bad set $V'$:

$$\Pr[\exists \text{ bad set}] \leq \left(\frac{n}{n/2^{\omega+1}}\right)^{(1-\omega)\delta+\omega} \cdot 2^{\left(\frac{n}{n/2^{\omega+1}}\right)^{(1-\omega)\delta+\omega}}$$

$$\leq \left(\frac{n}{n/2^{\omega+1}}\right)^{\frac{n}{n/2^{\omega+1}}} \cdot 2^{\left(\frac{n}{n/2^{\omega+1}}\right)^{(1-\omega)\delta+\omega}}$$

$$\leq (e \cdot 2^{\omega+1})^{\frac{n}{n/2^{\omega+1}}} \cdot 2^{\left(\frac{n}{n/2^{\omega+1}}\right)^{(1-\omega)\delta+\omega}}$$

$$\leq \left[(e \cdot 2^{\omega+1}) \cdot 2^{\left(\frac{n}{n/2^{\omega+1}}\right)^{(1-\omega)\delta+\omega}}\right]^{\frac{n}{n/2^{\omega+1}}}$$

Here, we used the standard inequality $\left(\frac{n}{k}\right)^{k} \leq (\frac{en}{k})^{k}$. If we select $\delta$ such that:

$$\frac{2\omega + \log_2 e + 1}{\omega - 1} < \delta,$$

then we have that the probability that there exists a bad set is at most $c_{29}(\delta, \omega)^n$ for

$$c_{29}(\delta, \omega) = 2^{\left(\frac{n}{n/2^{\omega+1}}\right)^{(1-\omega)\delta+\omega+\log_2 e + 1}}.$$

This proves the lemma.

\textbf{Lemma 30} Let $G_{n,d}$ be a depth-$d$ stack of $\delta$-random sandwich graphs with $n$ vertices on each level. Then for any integer $n_0 > 0$,

- every sandwich graph in the stack is an $(m, m, n/4)$-consecutively avoiding graph, for all $n_0 \leq m \leq 3n/4$, and
- every sandwich graph in the stack is $(n/2^{\omega+1}, n/2^{\omega})$-everywhere avoiding,

except with probability:

$$p_{\text{fail}}(\delta, \omega, n_0, n, d) \leq 2nd \cdot c_{28}(\delta, 2)^{n_0} + d \cdot c_{29}(\delta, \omega)^n.$$

\textit{Proof.} The probability that $G_{n,d}$ does not satisfy the first property is at most $p_1 \leq 2nd \cdot c_{28}(\delta, 2)^{n_0}$ by Lemma 28 and a Union Bound over the $d$ levels of the graph. The probability that $G_{n,d}$ does not satisfy the second property is at most $p_2 \leq d \cdot c_{29}(\delta, \omega)^n$ by Lemma 29 and an application of the Union Bound over the $d$ levels of the graph. The probability that either bad event occurs is at most $p_{\text{fail}} \leq p_1 + p_2$, by the Union Bound.

\textbf{Lemma 31} Let $G_{n,d}$ be a depth-$d$ stack of $\delta$-random sandwich graphs with $n$ vertices on each level. Then, pebbling the topologically last vertex of $G_{n,d}$ with at most $S$ pebbles requires time $T$ such that:

(a) $S \cdot T \geq dn^2/8$ for space usage $n_0 \leq S$, and
(b) $S \cdot T \geq (2^d - 1)n^2/8$ for space usage $n_0 \leq S < n/2^{\omega+1}$,

except with probability $p_{\text{fail}}(\delta, \omega, n_0, n, d)$, defined in Lemma 30.
More formally, we can divide any sequence of pebbling moves we must first place a pebble on the last vertex of level induction hypothesis, this requires at least $n^2/8$ pebbling moves. Thus $S \cdot T \geq n^2/8$. Since for $S \geq n/8$, $T \geq n$, we have that $S \cdot T \geq n^2/8$ holds for all $S \geq n_0$.

**Induction Step.** Assume the theorem holds for stacks of depth at most $d - 1$. At the start of a pebbling of $G_{n,d}$, there are no pebbles on the graph.

To prove Part (a): To place a pebble the first pebble on level $d$ of the graph, we must first place a pebble on the last vertex of level $d - 1$ of the graph. By the induction hypothesis, this requires at least $(d - 1)n^2/(8S)$ pebbling moves. Now there are no pebbles on the last level of the graph, by Corollary 24, pebbling an $(S, S, n/4)$-consecutively avoiding graph with $n_0 \leq S \leq 3n/4$ pebbles requires at least $n^3/8$ pebbling moves. Thus $T \geq n^3/8$, and Part (a) is proved.

To prove Part (b): To place a pebble the first pebble on the last level of the graph, we must first place a pebble on the last vertex of level $d - 1$ of the graph. By the induction hypothesis, this requires at least $T_{d-1} \geq (2^{d-1} - 1)n^2/(8S)$ pebbling moves, if $S < n/2^{\omega+1}$.

By Corollary 24, pebbling an $(S, S, n/4)$-consecutively avoiding graph that has no pebbles on the bottom half of the graph with $n_0 \leq S \leq 3n/4$ pebbles requires pebbling at least $n/(2S)$ subsets of $n/4$ vertices on the “top half” of the sandwich graph.

These $n/4$ vertices are the bottom vertices of a depth-$(d-1)$ stack of $(n/2^{\omega+1}, n/2^\omega)$-everywhere avoiding graphs. By Lemma 25, pebbling any $n/2^{\omega+1}$ unpebbled vertices of a depth-$(d - 1)$ stack of such graphs with at most $n/2^{\omega+1}$ pebbles requires at least $2^{d-1}(n/2^{\omega+1})$ moves. Since there are $(n/4)/(n/2^{\omega+1}) = 2^{\omega+1}/4 = 2^{\omega-1}$ segments of such vertices, the total time required to pebble them is $2^{\omega-1}(2^{d-1}n/2^{\omega+1}) = 2^d n/8$.\(^{20}\)

The total time to pebble each of the $n/(2S)$ segments of such vertices is then:

\[
T_d \geq \frac{n}{2S} \cdot \frac{2^d n}{8} = \frac{2^{d-1}n^2}{8S}.
\]

More formally, we can divide any sequence of pebbling moves $\mathcal{M}$ that pebbles all of these $2^{\omega-1}$ segments into $2^{\omega-1}$ distinct sub-sequences of consecutive pebbling moves. By Lemma 25, each of these sub-sequences must contain at least $2^{d-1}n/2^{\omega+1}$ moves.
Fig. 12: Components of the data-dependency graph for one Balloon mixing round. Here, $v_i(t)$ represents the value stored in the $i$th block in the main memory buffer at the $t$th mixing round.

So the total time to place a pebble on the last vertex of the $d$-th level of the graph is:

$$T \geq T_d + T_{d-1} \geq \frac{2^{d-1}n^2}{8S} + \frac{(2^{d-1} - 1)n^2}{8S} = \frac{(2^d - 1)n^2}{8S},$$

and Part (b) is proved.

\[\square\]

### E From Pebbling to Memory-Hardness

In this section, we complete the security analysis of the Balloon function. We present the formal proof of the claim that the space $S$ and time $T$ required to compute the $r$-round Balloon function satisfies (roughly) $S \cdot T \geq rn^2/8$. The other pieces of Informal Theorem 1 follow from repeated application of the same technique.

**Claim 32** The output of the $n$-block $r$-round Balloon construction is the labeling, in the sense of Definition 3, of a depth-$r$ stack of localized and degree-reduced $\delta$-random sandwich graphs.

**Proof.** Follows by inspection of the Balloon algorithm (Figure 1).

\[\square\]

**Theorem 33 (Formal statement of first part of Informal Theorem 1)**

Let $k$ denote the block size (in bits) of the underlying cryptographic hash function used in the Balloon constructions. Fix an integer $n_0 > 0$. Any algorithm $A$ that computes the output of the $n$-block $r$-round Balloon construction (with security parameter $\delta = 7$), makes $T$ random-oracle queries, and uses fewer than $\sigma$ bits of storage space, such that

$$T < \frac{rn^2}{8S^*} \quad \text{and} \quad \sigma < S^* \left( k - \log_2 \left( \frac{rn^2}{8S^*} \right) \right) - k,$$

for some $n_0 \leq S^* < 3n/4$, then the probability that $A$ succeeds is at most

$$\frac{T + 1}{2^k} + p_{\text{fail}}(7, n_0, n, r),$$

49
where \( p_{\text{fail}}(\cdot, \cdot, \cdot, \cdot) \) is defined in Lemma 30.

To make the failure probability concrete, we give an example: if the adversary can make \( T = 2^{60} \) random-oracle queries, we use a random oracle with a \( k = 512 \)-bit output size, we fix \( n_0 = 128 \), we use a buffer with \( n = 2^{14} \) blocks, and we run the Balloon hashing algorithm for \( r = 8 \) rounds, then adversary \( \mathcal{A} \)'s success probability in Theorem 33 is much smaller than \( 2^{-256} \).

**Proof of Theorem 33.** Fix an algorithm \( \mathcal{A} \). By Claim 32, \( \mathcal{A} \) outputs the labeling of a depth-\( r \) stack of \( \delta \)-random sandwich graphs. By Lemma 31, shows that pebbling these graphs with \( S \geq n_0 \) pebbles takes time at least \( n^2/(8S) \), except with some small probability. Let \( B \) denote this event that the pebbling bound does not hold.

Then we have:

\[
\Pr[\mathcal{A} \text{ succeeds}] = \Pr[\mathcal{A} \text{ succeeds}|B] \cdot \Pr[B] + \Pr[\mathcal{A} \text{ succeeds}|\neg B] \cdot \Pr[\neg B] \\
\leq \Pr[\mathcal{A} \text{ succeeds}|\neg B] + \Pr[B].
\]

From Lemma 31, \( \Pr[B] \leq p_{\text{fail}}(\delta, 2, n_0, n, r) \).

Conditioned on \( \neg B \), we may use Lemma 31 in conjunction with Theorem 5 to show that \( \Pr[\mathcal{A} \text{ succeeds}|\neg B] \) is small. In particular, for any \( S^* \geq n_0 \), there does not exist a pebbling of the graph that uses than \( T^* = \frac{n^2}{8S^*} \) pebbling moves.\(^{21}\)

We then apply Theorem 5 to conclude that \( \Pr[\mathcal{A} \text{ succeeds}|\neg B] \leq (T + 1)/2^k \).

This completes the proof.

The other parts of Informal Theorem 1 follow from a similar analysis.

## F Argon2i Proof of Security

In this section, we show that the proof techniques we introduce for the analysis of the Balloon algorithm can also apply to prove the first known memory-hardness results on the Argon2i algorithm.

We focus here on the single-pass variant of Argon2i, described in Section 5, and do not attempt to generalize our results to the multi-pass variant. As in Section 5, we analyze an idealized version of Argon2i, in which the randomly chosen predecessor of a vertex is chosen using the uniform distribution over the topologically prior vertices.

**Theorem 34** Let \( A_n \) denote the single-pass Argon2i data-dependency graph on \( n \) blocks. Fix a positive integer \( n_0 \). Then pebbling \( A_n \) with \( n_0 < S < n/24 \) pebbles requires time \( T \) such that:

\[
S \cdot T \geq \frac{n^2}{192}.
\]

\(^{21}\) The graph used in the Balloon construction is degree-reduced and localized, so it is not precisely a \( \delta \)-random sandwich graph, but by Remark 26, the pebbling time-space lower bounds still apply to the degree-reduced and localized graph.
except with probability \( n^2 2^{-n_0 - 1} \).

Note that the memory-hardness theorem that we are able to prove here about single-pass Argon2i is much weaker than the corresponding theorem we can prove about the Balloon algorithm (Informal Theorem 1). Essentially, this theorem says that an attacker computing single-pass Argon2i cannot save more than a factor of \( 192 \times \) in space without having to pay with some increase in computation cost.

Using the language of Section 6.1, we find that an attacker’s advantage at computing single-pass Argon2i is bounded by

\[
\text{Adv}_{\mathcal{S}, \text{Argon2i}(r=1)} = \frac{n^2}{n^2/192} = 192,
\]

for space usage \( S \) satisfying \( n_0 < S < n/24 \), whereas for Catena-BRG the advantage is at most 32 and for single-round Balloon Hashing, the advantage is at most 16.

For the purposes of the security analysis, it will be more convenient to look at the graph \( A_{2n} \). Take the graph \( A_{2n} \) with and write its vertex set in topological order as: \((u_1, \ldots, u_n, v_1, \ldots, v_n)\). We can now think of the \( u \) vertices as the “top” vertices of the graph, and the \( v \) vertices as the “bottom” vertices in the graph.

Then we have the following claim:

**Claim 35** Consider any set of \( 12m \) bottom vertices of the graph \( A_{2n} \). These vertices are an \((m, n/4)\)-avoiding set, in the sense of Definition 11, except with probability \( 2^{1-m} \).

**Proof.** Let \( v_i \) be some bottom vertex in \( A_{2n} \). The vertex \( v_i \) has one predecessor (call it \( \text{pred}(v_i) \)) chosen i.u.r. from the set \( \{u_1, \ldots, u_n, v_1, \ldots, v_{i-1}\} \). Conditioned on the event that that \( \text{pred}(v_i) \in \{u_1, \ldots, u_n\} \), this predecessor is chosen i.u.r. from the set of top vertices \( \{u_1, \ldots, u_n\} \). Let \( T_i \) be an indicator random variable taking on value “1” when vertex \( v_i \)’s randomly chosen predecessor is a top vertex. Then \( \Pr[T_i = 1] \geq 1/2 \).

Fix a set \( V' \) of \( \kappa m \) vertices on the bottom half of the graph. We want to show that, with high probability, \( V' \) has at least \( 3m \) predecessors in the top-half of the graph. The expected number of top-half predecessors of the set \( V' \) is at least \( \kappa m/2 \), and we want to bound the probability that there are at most \( 3m \) top-half predecessors. A standard Chernoff bound gives that, for \( \kappa m \) independent Poisson trials with success probability at least \( 1/2 \), the probability that fewer than \( 3m \) of them succeed is bounded by

\[
\Pr \left[ \sum_{v_i \in V'} T_i < 3m \right] < \exp \left( -\frac{\kappa m}{2} \right) \left( \frac{1 - \frac{3}{2}}{2} \right) \leq 2^{-m},
\]

for \( \kappa > 6 \). Now taking \( \kappa = 12 \) gives:

\[
\Pr \left[ \sum_{v_i \in V'} T_i < 3m \right] < \exp \left( -\frac{3m}{4} \right) < 2^{-m}.
\]

51
Now, by Lemma 7 (with $\delta = 3$ and $\omega = 2$), the probability that the predecessors of set $V'$ are not $(m, n/4)$-well-spread over the top vertices is at most $2^{-m}$. By the Union Bound, the probability that either $V'$ has fewer than $3m$ i.u.r. top-half predecessors or that these predecessors are poorly spread is at most $2 \cdot 2^{-m} = 2^{1-m}$. By Lemma 12, the set $V'$ is then an $(m, n/4)$-avoiding set, except with probability $2^{1-m}$. \hfill $\square$

Proof of Theorem 34. Fix an integer $n_0$. We show that $A_{2n}$ is a $(12m, m, n/4)$-consecutively avoiding graph when $n_0 \leq m \leq n/12$, except with probability $n^2/48$. The probability that any consecutive set of $12m$ vertices on the bottom level of the graph is not an $(m, n/4)$-avoiding set is at most $\sum_{i=1}^{n} \sum_{m=128}^{2^{1-m}} 2^{1-m} \leq 2^{n^2/48}$, using Claim 35 and the Union Bound.

Now, we apply Corollary 24 (with $\kappa = 12m$, $\sigma = m$, $\rho = n/4$) to conclude that pebbling $A_{2n}$ with at most $S$ pebbles requires at least $T \geq n^2/48$ pebbling moves, when $n_0 < S < n/12$, except with probability $2^{n^2/48}$. Alternatively, we can write: $S \cdot T \geq n^2/48$.

Now, since we are interested in the complexity of pebbling $A_n$ (not $A_{2n}$), we can divide the $n$s by two to find that pebbling $A_n$ with at most $S$ pebbles requires $S \cdot T \geq n^2/192$ pebbling moves, when $n_0 < S < n/24$, except with probability $n^2/2^{n_0-1}$. \hfill $\square$

We can convert the pebbling lower bound into a time-space lower bound in the random-oracle model using the techniques of Appendix E.

G Analysis of Scrypt

In this section, we show that the proof techniques we have used to analyze the memory-hardness of the Balloon algorithm are useful for analyzing scrypt. In particular, we give a very simple proof that a simplified version of the core routine in the scrypt password hashing algorithm ("ROMix") is memory-hard in the random-oracle model [65]. Although scrypt uses a password-dependent access pattern, we show that even if scrypt used a password-independent access pattern, it would still be a memory-hard function in the sense of Section 2.2.

Alwen et al. [4] give a proof memory-hardness of scrypt in the stronger "parallel random-oracle model" (Section 4) under combinatorial conjectures (see Footnote 8). Here we prove memory-hardness in the simpler, and less powerful, single-instance sequential model of computation. At the same time, our techniques yield an unconditional result and a tighter time-space lower bound $\Omega(n^2)$, rather than the $\Omega(n^2/\log^2 n)$ bound of Alwen et al. [4].

Our goal is not to prove a more powerful statement about scrypt—just to show that our techniques are easy to apply and are broadly relevant to the analysis of other memory-hard functions.

Simplified scrypt. The simplified variant of scrypt we consider here—which operates on a buffer $(x_1, \ldots, x_n)$ of $n$ memory blocks using random oracles $H_1$, $H_2$, and $H_3$—operates as follows:
1. Set $x_1 \leftarrow H_1(1, \text{passwd}, \text{salt})$.
2. For $i = 2, \ldots, n$: Set $x_i \leftarrow H_1(i, x_{i-1})$.
3. Set $y \leftarrow x_n$.
4. For $i = 2, \ldots, n$:
   - Set $r \leftarrow H_2(i, \text{salt}) \in \{1, \ldots, n\}$.
   - Set $y \leftarrow H_3(i, y, x_r)$.
5. Output $y$.

This variant of scrypt uses a password-independent access pattern, so our time-space lower bound applies not only to conventional scrypt but also to this cache-attack-safe scrypt variant. In Step 4 above, if we selected $r$ by also hashing in the value of $y$, then we would have a hash function with a password-dependent scheme that behaves more like traditional scrypt.

**Theorem 36** Fix a positive integer $n_0$. Let $G_n$ denote the data-dependency graph for the core scrypt function ($\text{ROMix}$) on $n$ memory blocks. Then any strategy for pebbling $G_n$ using at most $S$ pebbles requires at least $T$ pebbling moves, such that:

$$S \cdot T \geq \frac{n^2}{24},$$

for $n_0 \leq S < n/4$, with probability $n^2 2^{-n_0}$.

**Proof.** By inspection of the ROMix algorithm, one sees that $G_n$ is a 1-random sandwich graph with $n$ top and $n$ bottom vertices.

We now use an argument similar to that of Lemma 28 to show that $G_n$ is hard to pebble with few pebbles. Fix an integer $m$ such that $128 \leq m \leq n/4$. Then any set of $3m$ vertices on the bottom half of $G_n$ have $3m$ predecessors on the top half of $G_n$, chosen independently and uniformly at random (using the random oracle). By Lemma 7 (applied with $\delta = 3$ and $\omega = 2$), these predecessors are $(m, n/4)$-well spread over the top vertices, except with probability $2^{-m}$.

Fix an integer $n_0$. Using a Union Bound, the probability that any set of $3m$ consecutive vertices for $m \geq n_0$ induces a poorly spread set of predecessors is at most: $\sum_{i=1}^{n} \sum_{m=n_0}^{n/4} 2^{-m} \leq n^2 2^{-n_0}$. So, except with probability $n^2 2^{-n_0}$, the graph $G_n$ is $(3m, m, n/4)$-consecutively avoiding for $m \geq n_0$. Now we can apply Corollary 24 (with $\kappa = 3S$, $\sigma = S$, and $\rho = n/4$) to conclude that pebbling $G_n$ with at most $S$ pebbles, for $n_0 \leq S \leq n/4$, requires at least $T \geq \frac{n^2}{24S}$ pebbling moves with high probability. Thus, we have $S \cdot T \geq n^2/24$ when $n_0 \leq S \leq n/4$.

As in Appendix E, we can now turn the pebbling lower-bound into a time-space lower bound in the random-oracle model. We omit the details, since the conversion is mechanical.