

# Password Hashing and Memory Hardness

ENEE 457

# Recall Password Hashing

- Store only the hash values of the passwords in a table in the clear.
  - If Server is compromised, hard to recover password values given hash values.
- To defeat “Rainbow Tables” we can use a salt when hashing the password.

# But how to defeat the “Brute Force” Attack?

- Recall, only around  $95^6 \approx 7 \times 10^{11}$  hash evaluations required to recover a single password using Brute Force Search.
- Solution:

Password Scrambler  $PS$ :

1. Given a password  $pass$ , computing  $PS(pass)$  should be “fast enough” for the user.
2. Computing  $PS(pass)$  should be “as slow as possible” without contradicting 1.
3. Given  $y = PS(pass)$  there must be no significantly faster way to test  $q$  password candidates than by actually computing  $PS$  on each candidate.

# What About Parallel Computation?

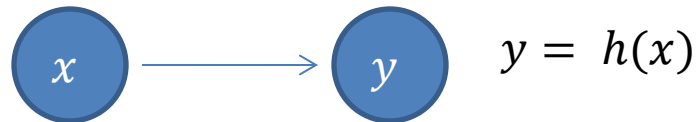
- Can't a  $b$ -core adversary always get a  $b$ -times speedup?
- Memory is expensive
  - Typical GPU or other cheap and massively-parallel hardware with lots of cores can only have a limited amount of fast (“cache”) memory for each single core
- Make the password scrambler PS not only intentionally slow on standard sequential computers, but also intentionally **memory-consuming**.
- Any adversary using  $b$  cores in parallel with less than about  $b$  times the memory of a sequential implementation must experience a strong slow-down.

# “Memory-Hard Functions”

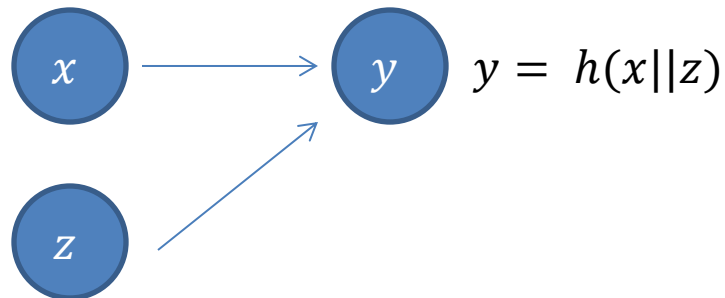
- Idea:
  - Start with an underlying hash function  $h$
  - Build a bigger hash function  $H$  from  $h$
- Assume to compute  $h(x_1, \dots, x_\ell)$  requires  $\ell$  units of time and  $\ell$  units of memory.

# Representing Hash Function Evaluation using a Graph

Start node  
corresponds to input



Each node corresponds to  
a value stored in memory.



Typically require **in-degree**  
to be **constant**, so that  
hash evaluation for each  
node takes constant time.

# (Parallel) Graph Pebbling

Let  $G = (V, E)$  be a DAG and  $T, S \subseteq V$  be node sets. Then a (legal) pebbling of  $G$  with starting configuration  $S$  and target  $T$  is a sequence  $(P_0, \dots, P_t)$  of subsets of  $V$  such that:

1.  $P_0 \subseteq S$
2. Pebbles are added only when their predecessors already have a pebble at the end of the previous step
3. At some point every target node is pebbled (not necessarily simultaneously).

We call a pebbling of  $G$  complete if  $S = \emptyset$  and  $T$  is the set of sink nodes of  $G$ .

# Space Complexity

Let  $G$  be a DAG,  $P = (P_0, \dots, P_t)$  be an arbitrary pebbling of  $G$  and  $\Pi$  be the set of all complete pebbblings of  $G$ . Then the (cumulative) cost of  $P$  and the cumulative complexity (CC) of  $G$  are defined respectively to be:

$$s\text{-cost}(P) := \max\{P_i : i \in \{0, \dots, t\}\}$$

$$sc(G) := \min\{s\text{-cost}(P) : P \in \Pi\}$$

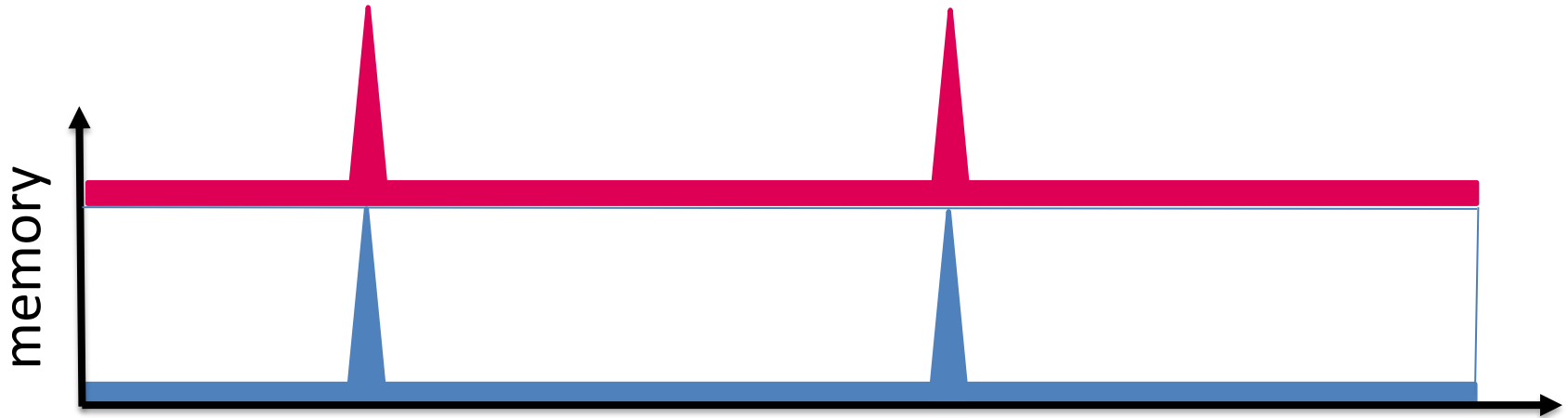
$$st\text{-cost}(P) := t \cdot \max\{P_i : i \in \{0, \dots, t\}\}$$

$$stc(G) := \min\{st\text{-cost}(P) : P \in \Pi\}$$

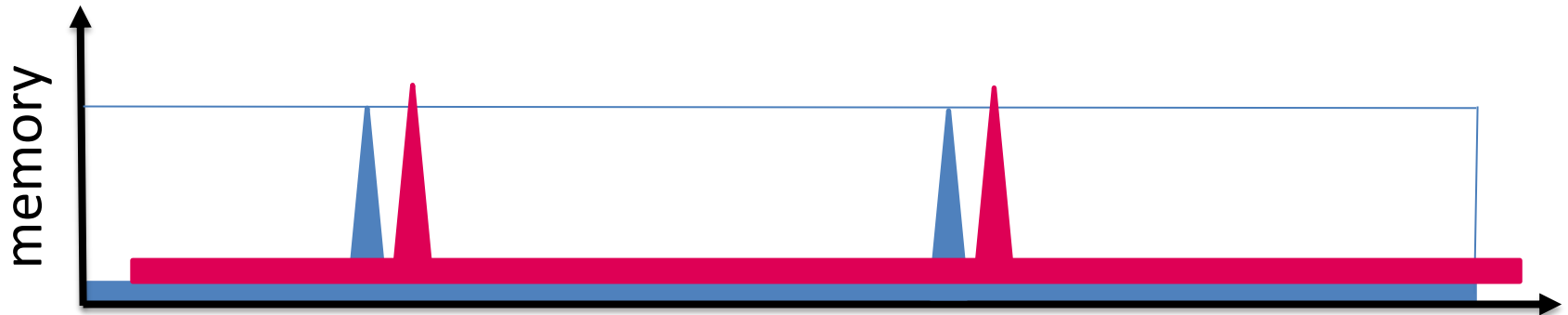


# Problem with Standard Notions

To compute two instances, a smart adversary won't do this!

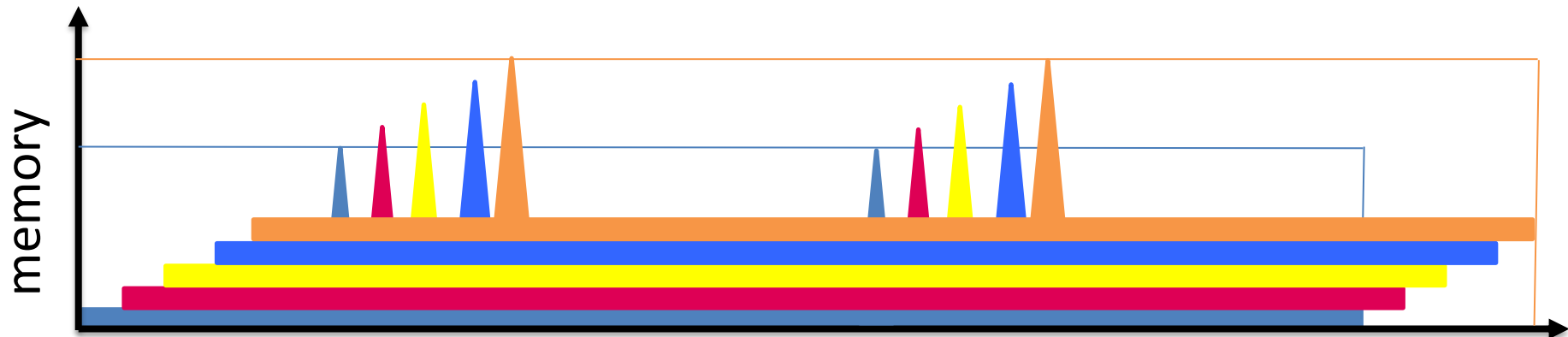


Instead:



# Problem with Standard Notions

Offset multiple computations by a little to keep cost low!



# Cumulative Pebbling Complexity

Let  $G$  be a DAG,  $P = (P_0, \dots, P_t)$  be an arbitrary pebbling of  $G$  and  $\Pi$  be the set of all complete peblings of  $G$ . Then the (cumulative) cost of  $P$  and the cumulative complexity (CC) of  $G$  are defined respectively to be:

$$p\text{-cost}(P) := \sum_{i=0}^t |P_i|$$

$$cc(G) := \min\{p\text{-cost}(P) : P \in \Pi\}$$

# Cumulative Pebbling Complexity

Lemma: Let  $G = G_1 + G_2$ . Then  $cc(G) = cc(G_1) + cc(G_2)$ .

Lemma: There exists a  $G$  such that  $stc(G^{\times n}) = O(stc(G))$ .

# Maximal CC?

Lemma: Let  $G$  be a DAG of size  $n$  and depth  $d$ .  
Then  $cc(G) \leq dn$ .

Maximal CC is at most  $n^2$  for an  $n$ -node graph.

Our focus: What is the Maximal CC we can achieve for graphs with constant in-degree?

# Case Study: Bit-Reversal Graphs

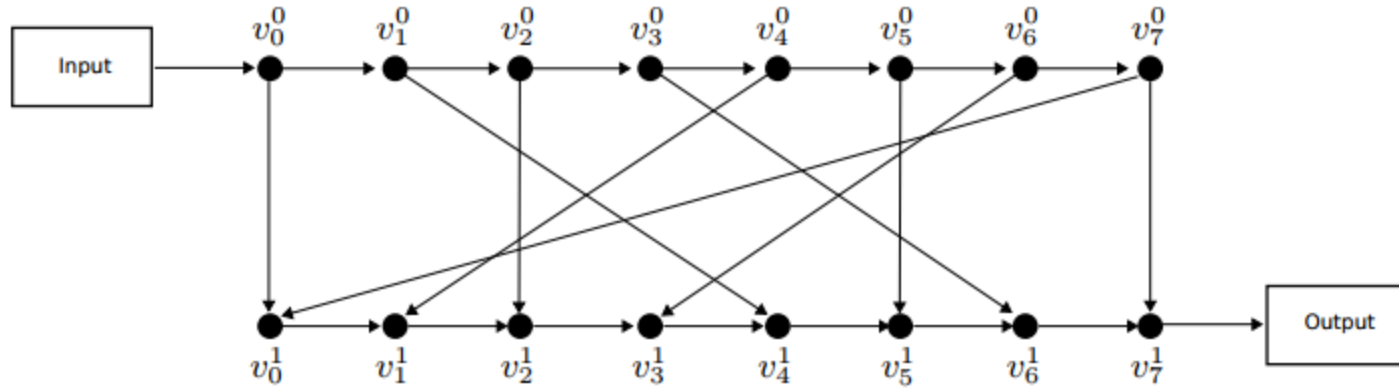


Fig. 1. An (8, 1)-BRG.

# Case Study: Bit-Reversal Graphs

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**Algorithm 2**  $(g, \lambda)$ -Bit-Reversal Hashing ( $BRH_\lambda^g$ )

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**Require:**  $g$  {Garlic},  $x$  {Value to Hash},  $\lambda$  {Depth},  $H$  {Hash Function}

**Ensure:**  $x$  {Password Hash}

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1:  $v_0 \leftarrow H(x)$ 
2: for  $i = 1, \dots, 2^g - 1$  do
3:    $v_i \leftarrow H(v_{i-1})$ 
4: end for
5: for  $k = 1, \dots, \lambda$  do
6:    $r_0 \leftarrow H(v_0 \parallel v_{2^g-1})$ 
7:   for  $i = 1, \dots, 2^g - 1$  do
8:      $r_i \leftarrow H(r_{i-1} \parallel v_{\tau(i)})$ 
9:   end for
10:   $v \leftarrow r$ 
11: end for
12: return  $r_{2^g-1}$ 
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# Case Study: Bit-Reversal Graphs

It was shown in [Lengauer Tarjan 82] that (in the sequential setting) any pebbling using  $S$  pebbles requires time  $T$  such that  $ST = O(n^2)$ .

Such graphs were suggested as candidates for password hashing. E.g. in the **Catena** framework (finalist in Password Hashing Competition [PHC]).

We will describe an algorithm which can pebble the bit-reversal graph of size  $n$  using cumulative cost of at most  $O(n^{1.5})$ .



# CC of Bit-Reversal Graphs?

Theorem: A Bit-Reversal graph  $G$  of size  $n$  has  $cc(G) = O(n^{1.5})$ .

Extends to any “sandwich” graph:

A chain of  $n$  nodes (numbered 1 through  $n$ ) with arbitrary additional edges connecting nodes from the first half of the chain with nodes of the second half of the chain such that no node has in-degree greater than 2.

# CC of Bit-Reversal Graphs?

Proof: Consider the following strategy:

1. If  $i \bmod \sqrt{n} = 0$  then place a pebble on node 1
2. For each pebble on a node  $v \in [n]$  place a pebble on node  $v + 1$
3. Remove any pebble on nodes  $\left\{ \binom{n}{2} + 1, \dots, n \right\}$  except the one on the highest valued node.
4. Let  $m$  be the highest valued node with a pebble on it. Remove any pebbles on nodes  $v \in [n/2]$  except if  $(i - v) \bmod \sqrt{n} = 0$  or if there is an edge  $(v, m + j)$  for some  $0 < j < \sqrt{n}$  and  $m + j > n/2$ .

# CC of Bit-Reversal Graphs?

Proof (cont'd).

Must show (1) the above strategy is a legal pebbling (2) at any time there are  $O(\sqrt{n})$  pebbles on the graph.

For (1), must show that this step is legal: For each pebble on a node  $v \in [n]$  place a pebble on node  $v + 1$ . Legal for first  $n/2$  nodes, but not necessarily second  $n/2$  nodes. Why?

Key: for second  $n/2$  nodes only the highest pebble on node  $m$  remains from previous round (due to Rule 3). Node  $m+1$  has at most one additional incoming edge from first  $n/2$  nodes. This node must be pebbled due to the fact that each node is touched every  $\sqrt{n}$  iterations and the second half of Rule 4.

For (2), at most one pebble on second  $n/2$  nodes (due to Rule 3). Due to first half of Rule 4, at most  $\sqrt{n}$  pebbles remain. Due to second half of Rule 4 and the fact that each of the second  $n/2$  nodes has in-degree at most 2, at most an additional  $2\sqrt{n}$  pebbles remain.

# Scrypt

- Initially introduced by Percival '09.
- Used in proofs-of-work schemes for cryptocurrencies.
- Inspired the design of one of the Password-hashing Competition's [PHC] winners, Argon2d.
- Similar structure to “sandwich” graph, but is data-dependent.
- The edges in the graph depend on the outcome of the hashed data.

# Script

- Input  $X$
- Output  $S_n$
- $X_0 = X$  and for  $i = 1, \dots, n - 1$ :  $X_i = h(X_{i-1})$
- $S_0 = h(X_{n-1})$  and for  $i = 1, \dots, n$ :  $S_i = h(S_{i-1} \oplus X_{S_{i-1} \bmod n})$

# Scrypt is Maximally Memory Hard

Theorem (Alwen et al.): The cumulative complexity of Scrypt is  $\Omega(n^2)$ .