**CS**11921 - **Fall** 2023

# Algorithm Design & Analysis

## Amortized Analysis

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# Motivation

Problem. Given an array $B[0 \ \ldots \ k-1]$ of bits, representing a number $n < 2^k$, increment $n$.

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Examples. `00000` —INCREMENT→ `00001`

$n = 0, \ k = 5$ $\quad$ $n = 1, \ k = 5$

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Examples. 00000 —INCREMENT→ 00001 —INCREMENT→ 00010

$n = 1, \ k = 5$ $\quad$ $n = 2, \ k = 5$

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Examples. 00000 —INCREMENT→ 00001 —INCREMENT→ 00010

```
INCREMENT(B[], k)
  i = k-1
```

start at the rightmost bit

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INCREMENT(B[], k)
    i = k-1
    while (B[i] == 1 and i >= 0):
        B[i] = 0
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```

keep flipping 1's to 0's until a 0 is reached

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flip to 1 flip to 0's

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Example. 1 0 1 0 1 0 0 1 1 1 1 1 1

ignore | flip to 1 | flip to 0's

**Q** What is the running time of function **INCREMENT**?
Choose the *best* answer.

Cost Model. Count the number of *bit flips*.

**A.** $O(1)$

**B.** $O(k)$

**C.** $O(\log n)$

**D.** $O(n)$

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```

Example. 1 0 1 0 1 0 **0** 1 1 1 1 1 1

ignore · flip to 1 · flip to 0's

**Q** What is the running time of function **INCREMENT**?
Choose the *best* answer.

Cost Model. Count the number of *bit flips*.

A. $O(1)$ ← incorrect

B. $O(k)$ ← too pessimistic!

C. $O(\log n)$

D. $O(n)$ ← too pessimistic!

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What is the running time for counting from `0` to `n` by calling function **`INCREMENT`** repeatedly on an array of $k$ bits initialized to `0`'s?

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Choose the *best* answer.

**A.** $O(n)$

**B.** $O(n \log n)$

**C.** $O(nk)$

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$O(n)$ — correct and tight bound!

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correct but too pessimistic!

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Choose the *best* answer.

$O(n)$ — correct and tight bound! ... *why*?

**B.** $O(n \log n)$

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correct but too pessimistic!

**D.** $O(k \log n)$ — incorrect!

Why is it pessimistic to say: $n$ calls to **`INCREMENT`** $\times O(\log n) = O(n \log n)$?

**Answer.** Because each call to **`INCREMENT`** does not do $O(\log n)$ bit flips!

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Problem. What is the total number of bit-flips performed when counting from `0` to `n` by calling function **`INCREMENT`** repeatedly on an array of $k$ bits initialized to `0`'s?

Example. Counting to `15`.

```
 0  0 0 0 0 0
 1  0 0 0 0 1
 2  0 0 0 1 0
 3  0 0 0 1 1
 4  0 0 1 0 0
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**In general.**

The total number of bit flips is:

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**Implication.**

Since **`INCREMENT`** is called $n$ times and the running time is $O(n)$ in total, the running time of each call to **`INCREMENT`** in the sequence of calls is $O(1)$ *on average!*

**Takeaway**

When analyzing the worst case running time of a *sequence* of operations, we can:

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Analyze the worst case running time of a single operation and then multiply it by the number of times the operation is performed.

Example. Running time of $n$ increments $= n \times O(\log n)$

Problem. Might *overestimate* the worst case running time.

**Takeaway**

When analyzing the worst case running time of a *sequence* of operations, we can:

1. Analyze the worst case running time of a single operation and then multiply it by the number of times the operation is performed.

Example. Running time of $n$ increments $= n \times O(\log n)$

Problem. Might *overestimate* the worst case running time.

OR

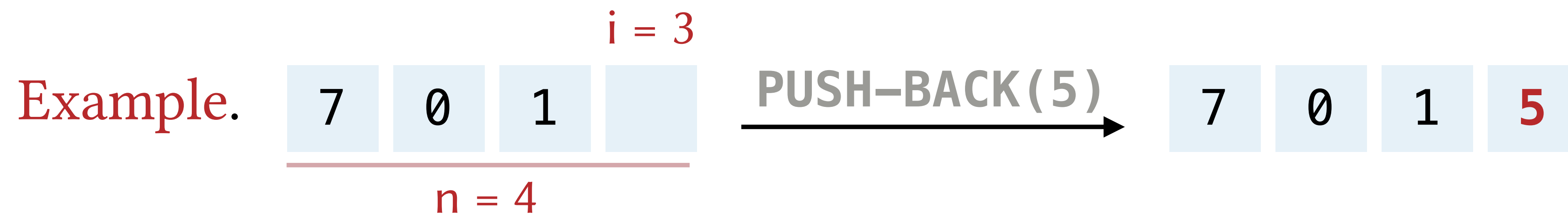
2. Reason about the total running time of the whole sequence of operations together.

Example. Incrementing $n$ times can't flip bits more than $2n$ times.

# Motivation

Problem. Given an array $A[]$ of size $n$, implement a **PUSH–BACK**(x) operation that inserts $x$ at the last vacant cell. If the number of occupied cells $i$ equals $n$, double the size of the array before inserting.

Example.



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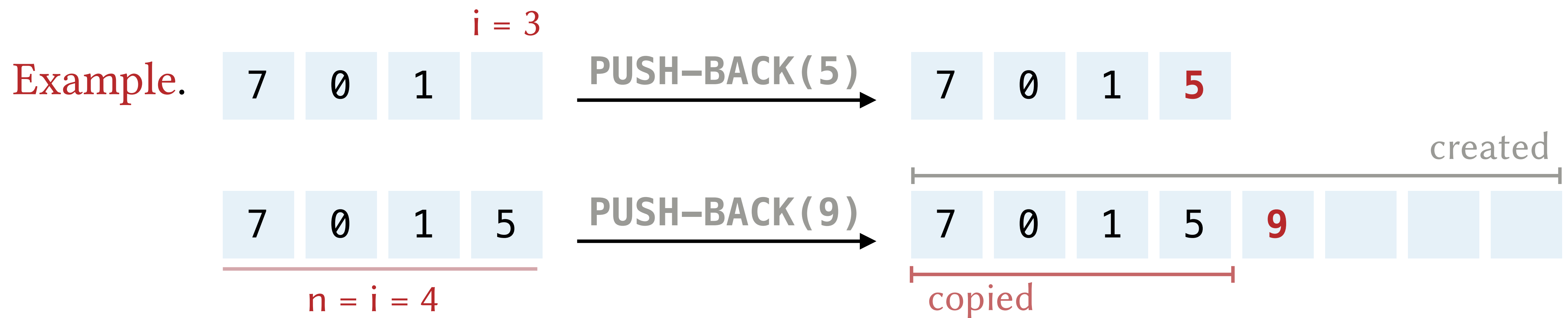

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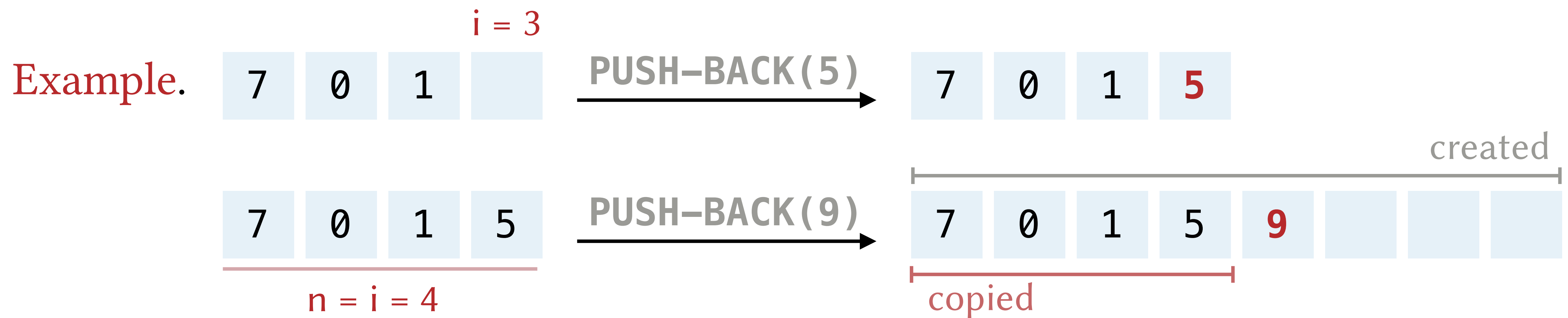


Assuming n and i are globally accessible.

```
PUSH-BACK(A[], x)
if (i < n): A[i] ← x
```

simply insert at index i if there is a vacant cell

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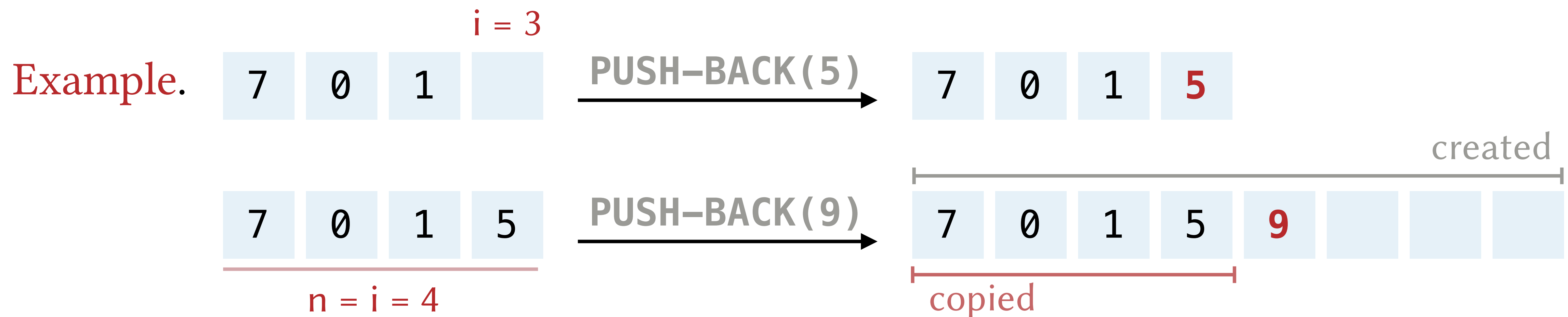


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if there are no vacant cells, double the size of the array

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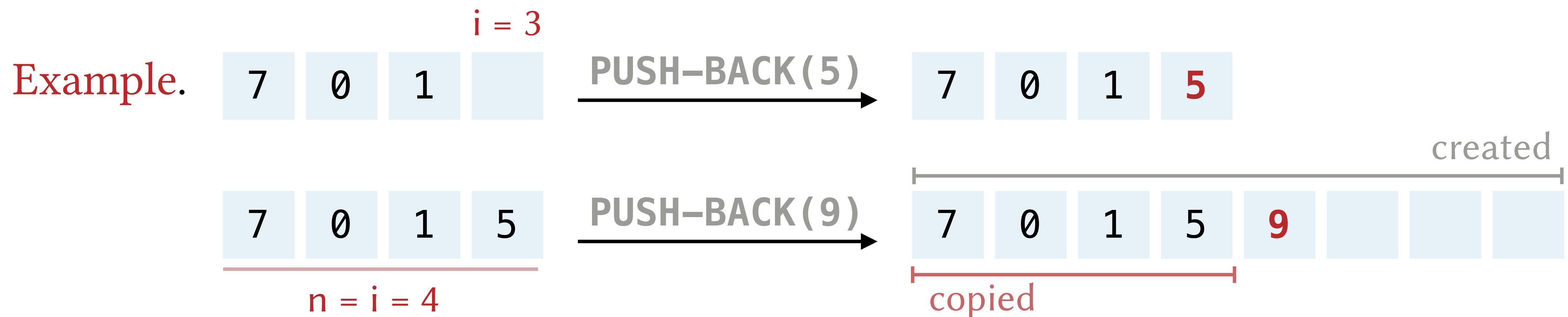


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insert in the newly created array

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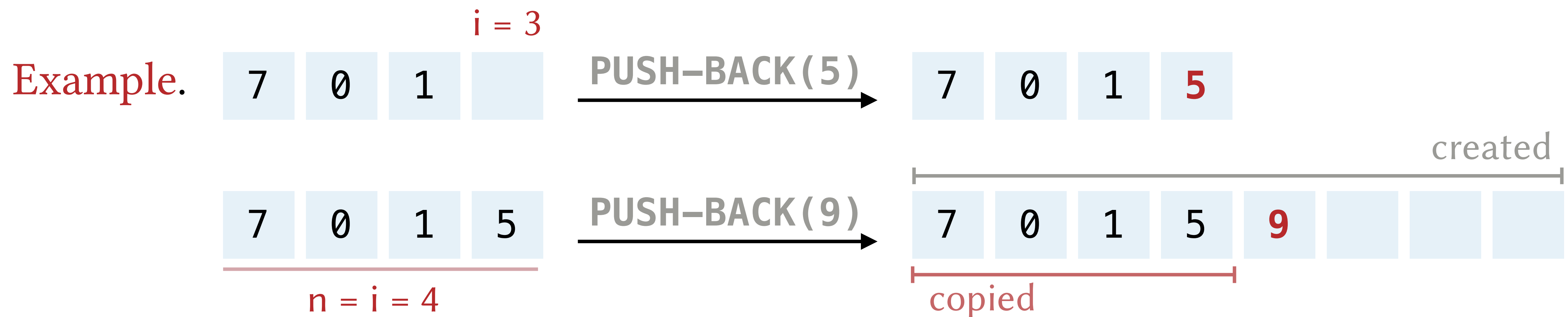


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**Q** What is the worst case running time of function **PUSH–BACK**?

Cost Model. Number of copied elements

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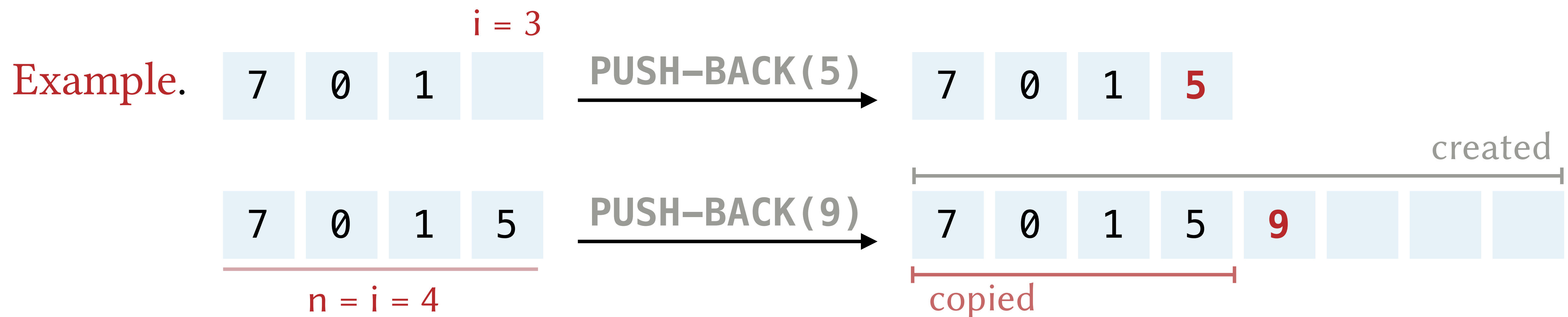


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i ← i+1
```

**Q** What is the worst case running time of function **PUSH-BACK**?

Cost Model. Number of copied elements

**A** $\Theta(n)$ because $n + 1$ elements are copied if the array is doubled

# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

Cost Model. Count the number of *element copies*.

Choose the *best* answer.

**A.** $O(n^2)$

**B.** $O(n \log n)$

**C.** $O(n)$

**D.** $O(\log n)$

# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

Cost Model. Count the number of *element copies*.

Choose the *best* answer.

**A.** $O(n^2)$

**B.** $O(n \log n)$

correct but too pessimistic!

$O(n)$

correct and tight bound!

**D.** $O(\log n)$

incorrect!

## Exercise

What is the worst case running time of calling **PUSH-BACK** $n$ times on a resizing array that is initially of size 1?

☒ 1

# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

| X | | 1 |
|---|---|---|
| X | X | 1 + 1 |

# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

| X | | | | 1 |
|---|---|---|---|---|
| X | X | | | 1 + 1 |
| X | X | X | | 1 + 2 |

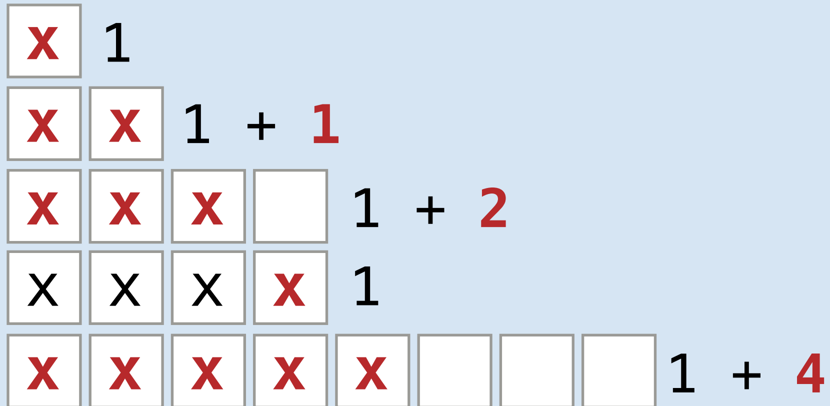
# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

| x | | | | 1 |
|---|---|---|---|---|
| x | x | | | 1 + 1 |
| x | x | x | | 1 + 2 |
| x | x | x | x | 1 |

# Exercise

What is the worst case running time of calling **PUSH-BACK** $n$ times on a resizing array that is initially of size 1?



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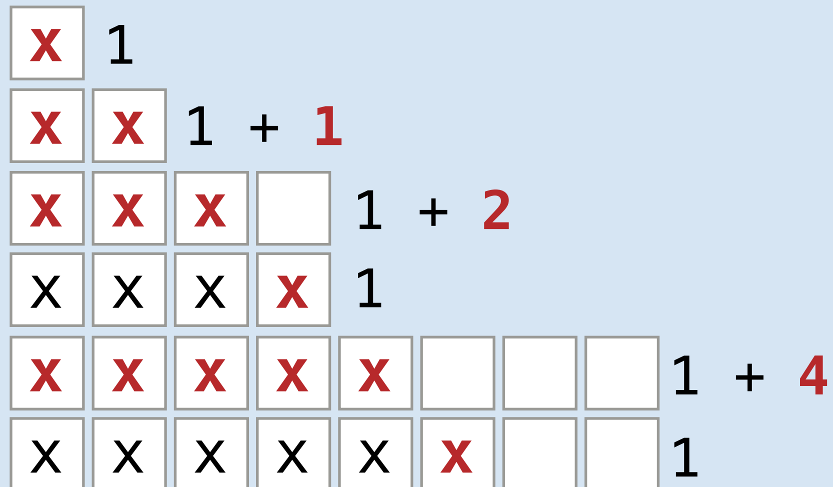
What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?



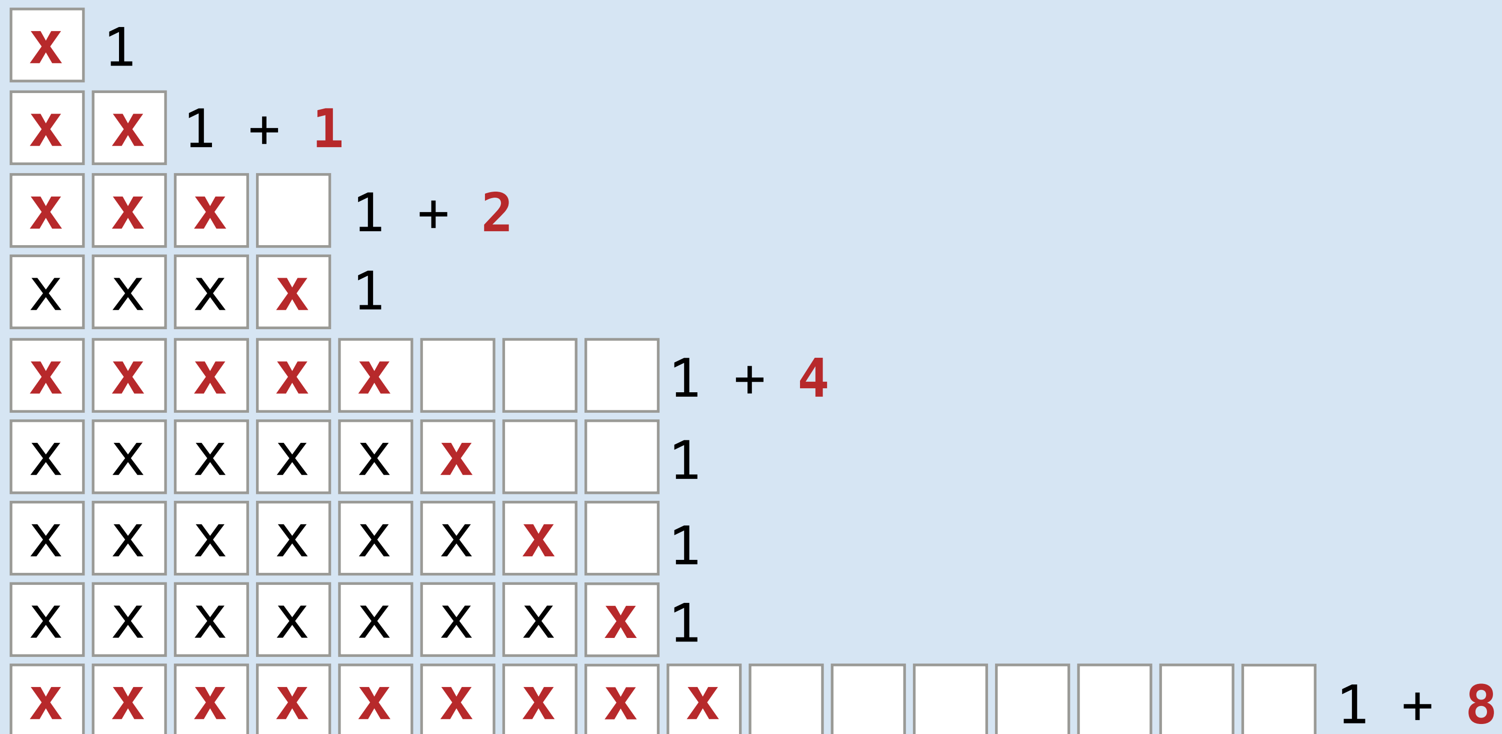
# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

| | | | | | | | | |
|---|---|---|---|---|---|---|---|---|
| x | 1 | | | | | | | |
| x | x | 1 + 1 | | | | | | |
| x | x | x | | 1 + 2 | | | | |
| x | x | x | x | 1 | | | | |
| x | x | x | x | x | | | | 1 + 4 |
| x | x | x | x | x | x | | | 1 |
| x | x | x | x | x | x | x | | 1 |
| x | x | x | x | x | x | x | x | 1 |

# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?



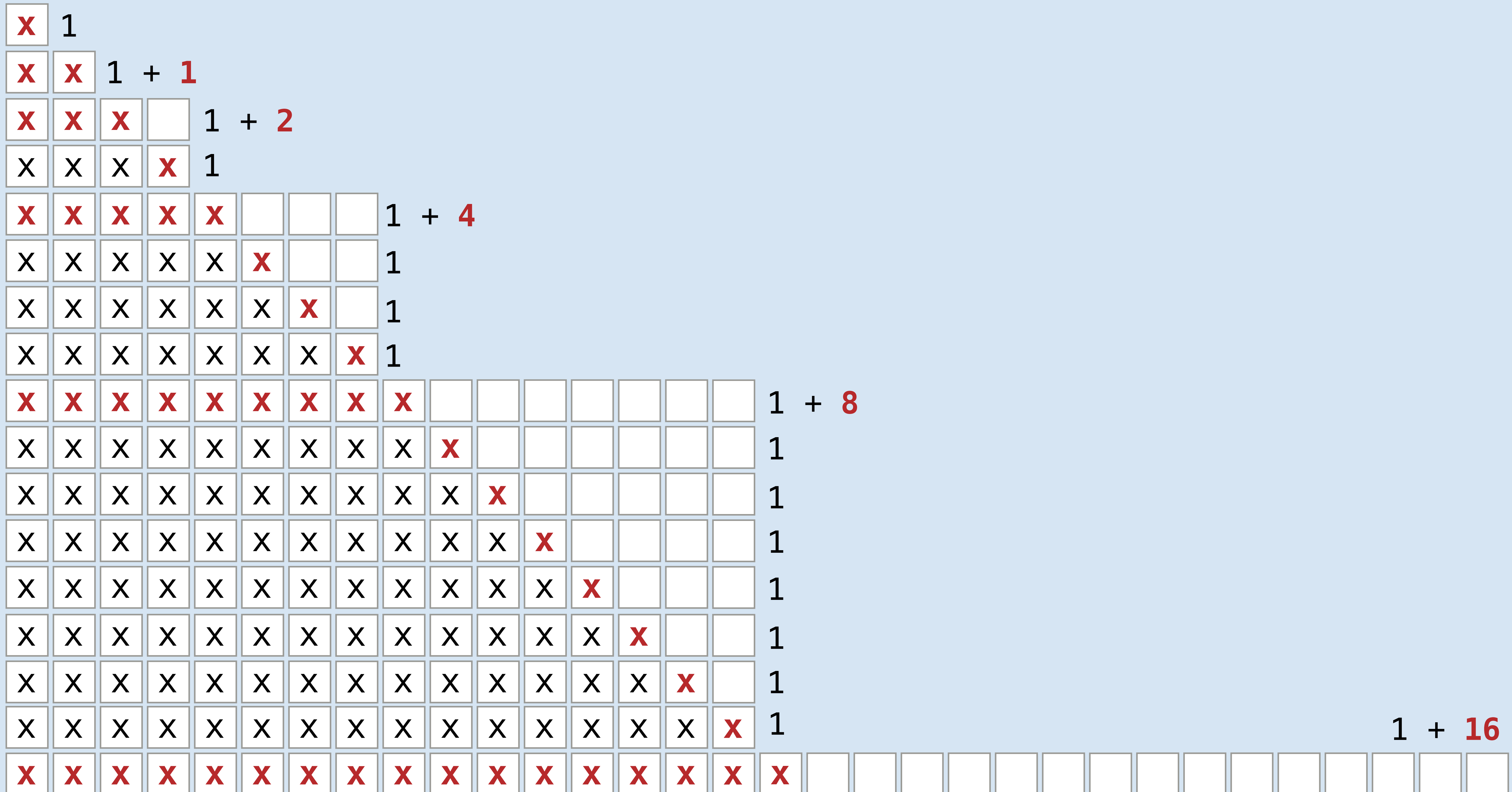
# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

| | | | | | | | | | | | | | | | | |
|---|---|---|---|---|---|---|---|---|---|---|---|---|---|---|---|---|
| x | | | | | | | | | | | | | | | | 1 |
| x | x | | | | | | | | | | | | | | | 1 + 1 |
| x | x | x | | | | | | | | | | | | | | 1 + 2 |
| x | x | x | x | | | | | | | | | | | | | 1 |
| x | x | x | x | x | | | | | | | | | | | | 1 + 4 |
| x | x | x | x | x | x | | | | | | | | | | | 1 |
| x | x | x | x | x | x | x | | | | | | | | | | 1 |
| x | x | x | x | x | x | x | x | | | | | | | | | 1 |
| x | x | x | x | x | x | x | x | x | | | | | | | | 1 + 8 |
| x | x | x | x | x | x | x | x | x | x | | | | | | | 1 |
| x | x | x | x | x | x | x | x | x | x | x | | | | | | 1 |
| x | x | x | x | x | x | x | x | x | x | x | x | | | | | 1 |
| x | x | x | x | x | x | x | x | x | x | x | x | x | | | | 1 |
| x | x | x | x | x | x | x | x | x | x | x | x | x | x | | | 1 |
| x | x | x | x | x | x | x | x | x | x | x | x | x | x | x | | 1 |
| x | x | x | x | x | x | x | x | x | x | x | x | x | x | x | x | 1 |

# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?



# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

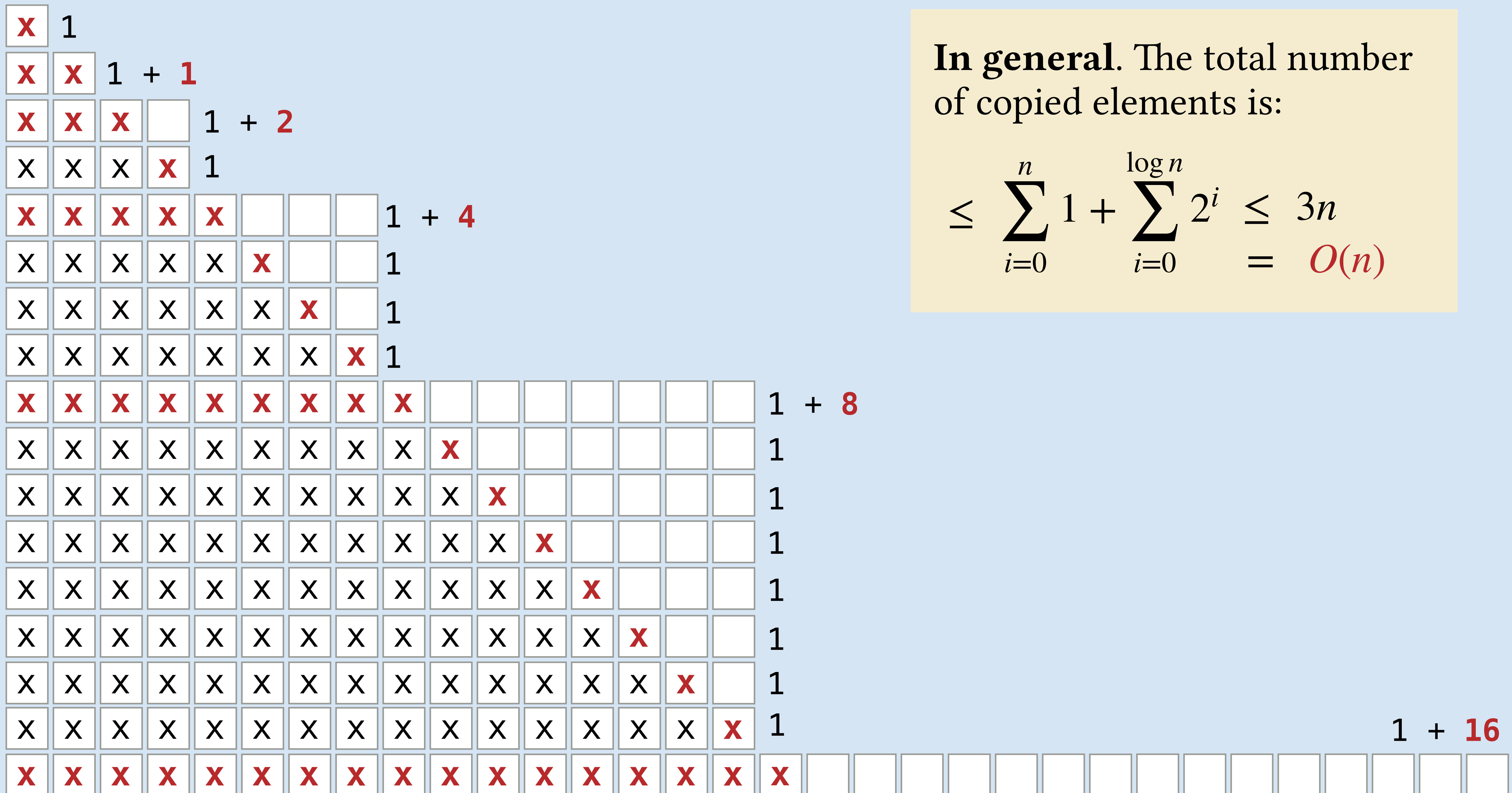
x 1
x x 1 + 1
x x x 1 + 2
x x x x 1
x x x x x 1 + 4
x x x x x x 1
x x x x x x x 1
x x x x x x x x 1
x x x x x x x x x 1 + 8
x x x x x x x x x x 1
x x x x x x x x x x x 1
x x x x x x x x x x x x 1
x x x x x x x x x x x x x 1
x x x x x x x x x x x x x x 1
x x x x x x x x x x x x x x x 1
x x x x x x x x x x x x x x x x 1
1 + 16
x x x x x x x x x x x x x x x x x

**In general**. The total number of copied elements is:

$$\leq \sum_{i=0}^{n} 1 + \sum_{i=0}^{\log n} 2^i$$

# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?

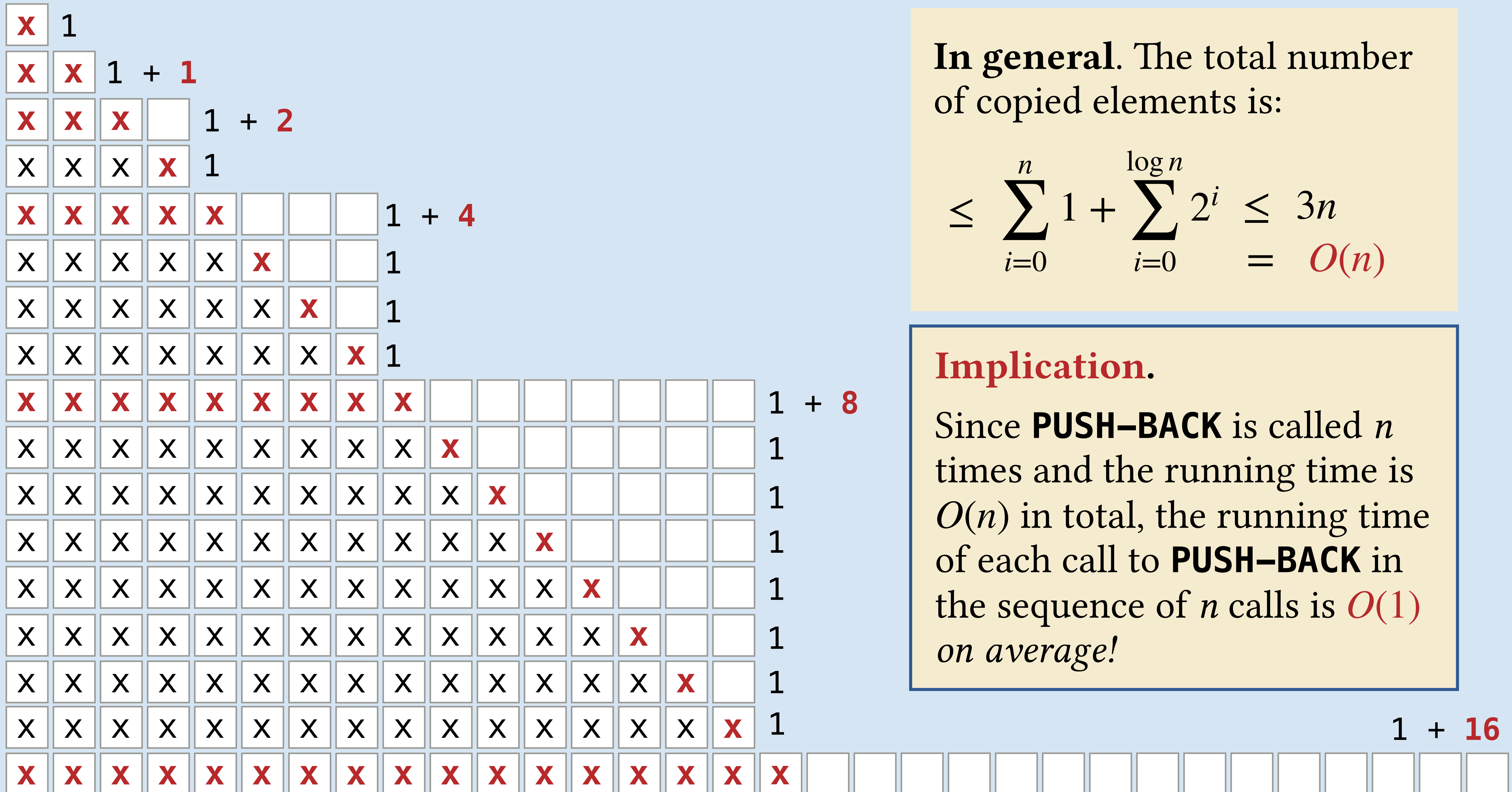



**In general.** The total number of copied elements is:

$$\leq \sum_{i=0}^{n} 1 + \sum_{i=0}^{\log n} 2^i \leq 3n = O(n)$$

# Exercise

What is the worst case running time of calling **PUSH–BACK** $n$ times on a resizing array that is initially of size 1?



*welcome to*

# Amortized Analysis

# Amortized Analysis

Goal. Analyze the worst case running time of a sequence of operations.

Worst Case Analysis

**PUSH-BACK** runs in $\Theta(n)$ in the worst case.

Interpretation. At least one of the cases can make the function run in $\Theta(n)$.

**INCREMENT** runs in $\Theta(\log n)$ in the worst case.

Interpretation. At least one of the cases can make the function run in $\Theta(\log n)$.

V.S.

Amortized Analysis

The running time of **PUSH-BACK** is $O(1)$ amortized.

The running time of **INCREMENT** is $O(1)$ amortized.

Interpretation. If we perform a sequence of operations, the running time overall will be in the order of $n$ in the worst case and each single operation will have performed a constant amount of work on average.

# Amortized Analysis

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Interpretation. If we perform a sequence of operations, the running time overall will be in the order of $n$ in the worst case and each single operation will have performed a constant amount of work on average.

Amortized analysis can be done in multiple ways. The method we used so far is called the **aggregate method**.

# **Amortized Analysis:** Accountant's Method

Idea. Use cheap frequent operations to pay for rare but expensive operations.

# Amortized Analysis: Accountant's Method

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- Assume that each unit of work costs `$1` and operation $i$ costs $c_i$ .

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- Assign a new cost $\hat{c}_i$ for operation $i$ (can be $<$, $>$ or $=$ to $c_i$ ).

# Amortized Analysis: Accountant's Method

Idea. Use cheap frequent operations to pay for rare but expensive operations.

- Assume that each unit of work costs `$1` and operation $i$ costs $c_i$.
- Assign a new cost $\hat{c}_i$ for operation $i$ (can be <, > or = to $c_i$).

- If $\hat{c}_i > c_i$ save the extra credit in the bank for use by other operations.
- If $\hat{c}_i < c_i$ consume from the credit stored in the bank.

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Goal. Show that credit always remains nonnegative, implying that $\sum \hat{c}_i \geq \sum c_i$

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---

Example. Array resizing.

**Actual Costs.**

- Copying a single element:
- Resizing the array:

**New Costs.**

- Copying a single element:
- Resizing the array:

# Amortized Analysis: Accountant's Method

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Example. Array resizing.

**Actual Costs.**

- Copying a single element: **`$1`**
- Resizing the array: **`$n`**
  ($n$ = number of elements added so far)

**New Costs.**

- Copying a single element:
- Resizing the array:

# Amortized Analysis: Accountant's Method

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---

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  ($n$ = number of elements added so far)

**New Costs.**

- Copying a single element: **`$2`**
- Resizing the array: **`$0`**

# Amortized Analysis: Accountant's Method

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---

Example. Array resizing.

**Actual Costs.**

- Copying a single element: **`$1`**
- Resizing the array: **`$n`**
  ($n$ = number of elements added so far)

**New Costs.**

- Copying a single element: **`$2`**
- Resizing the array: **`$0`**

We need to show that the bank credit will always remain nonnegative. I.e.
The total new cost is not less than (equal or worse than) the total actual cost.

# Accountants Method: Resizing Array Demo

**Actual Costs.**
- Copying a single element: $1
- Resizing the array: $n

**New Costs.**
- Copying a single element: $2
- Resizing the array: $0

**PUSH–BACK** use 1$ and save 1$ | $1 |

# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $2
- Resizing the array: $0

**PUSH-BACK** use 1$ and save 1$

**RESIZE** use 1$ from the saved credit to copy 1 element

# **Accountants Method:** Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $2
- Resizing the array: $0

**PUSH-BACK** use 1$ and save 1$ | $1 |

**RESIZE** use 1$ from the saved credit | | |

**PUSH-BACK** use 1$ and save 1$ | | $1 |

# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

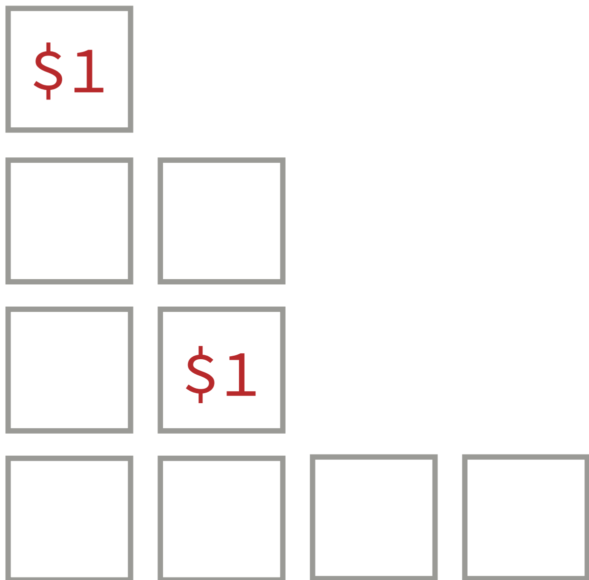
- Copying a single element: $2
- Resizing the array: $0

**PUSH–BACK** use 1$ and save 1$

**RESIZE** use 1$ from the saved credit

**PUSH–BACK** use 1$ and save 1$

**RESIZE** use 2$ from the saved credit to copy 2 elements



**OOPS!**

There is only $1 in the credit

The chosen new costs are bad!

# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: `$1`
- Resizing the array: `$n`

**New Costs.**

- Copying a single element: `$3`
- Resizing the array: `$0`

**SECOND ATTEMPT.** Use `$3` instead of `$2`

# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

**PUSH-BACK** use 1$ and save 2$

| $2 |
|---|

# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

**PUSH-BACK** use 1$ and save 2$

$2

**RESIZE** use 1$ from the saved credit to copy one element

| $1 | |
|---|---|

# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

**PUSH-BACK** use 1$ and save 2$ | $2 |

**RESIZE** use 1$ from the saved credit | $1 | |

**PUSH-BACK** use 1$ and save 2$ | $1 | $2 |

# **Accountants Method:** Resizing Array Demo

**Actual Costs.**
- Copying a single element: $1
- Resizing the array: $n

**New Costs.**
- Copying a single element: $3
- Resizing the array: $0

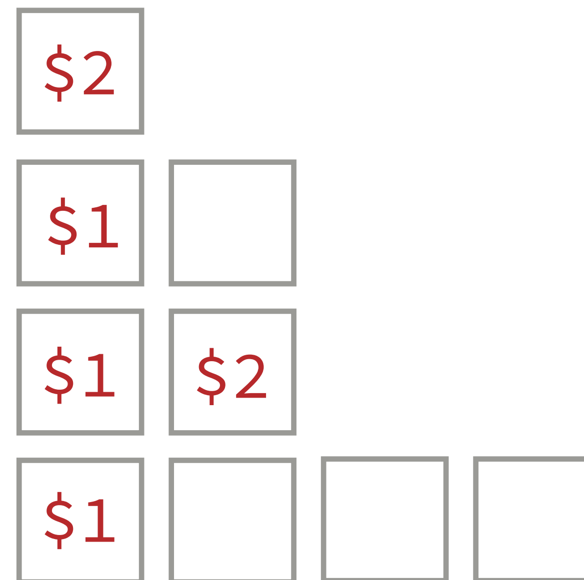
**SECOND ATTEMPT.** Use $3 instead of $2

**PUSH–BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

**PUSH–BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit to copy 2 elements



# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

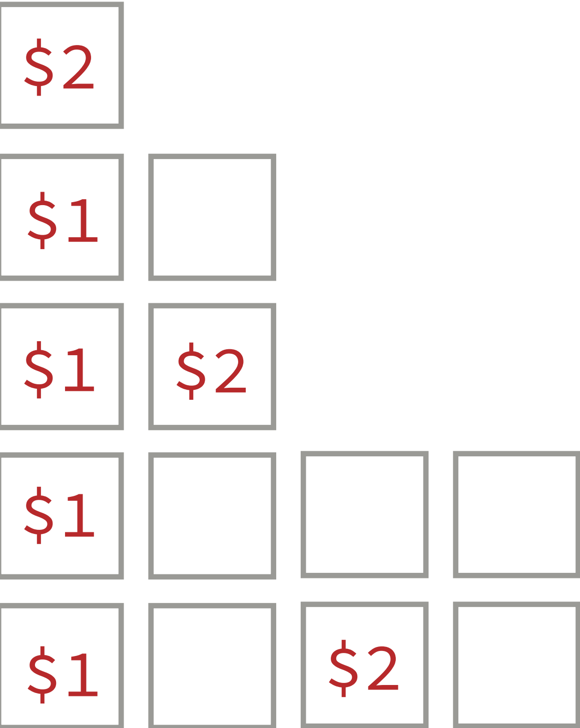
**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit

**PUSH-BACK** use 1$ and save 2$



# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

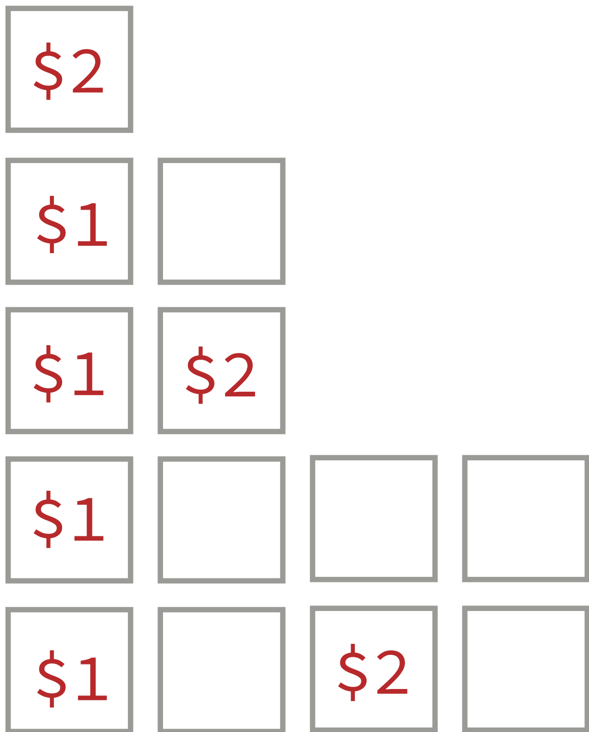
**PUSH–BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

**PUSH–BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit

**PUSH–BACK** 2 times: use 1$ and save 2$



# **Accountants Method:** Resizing Array Demo

**Actual Costs.**
- Copying a single element: `$1`
- Resizing the array: `$n`

**New Costs.**
- Copying a single element: `$3`
- Resizing the array: `$0`

**SECOND ATTEMPT.** Use `$3` instead of `$2`

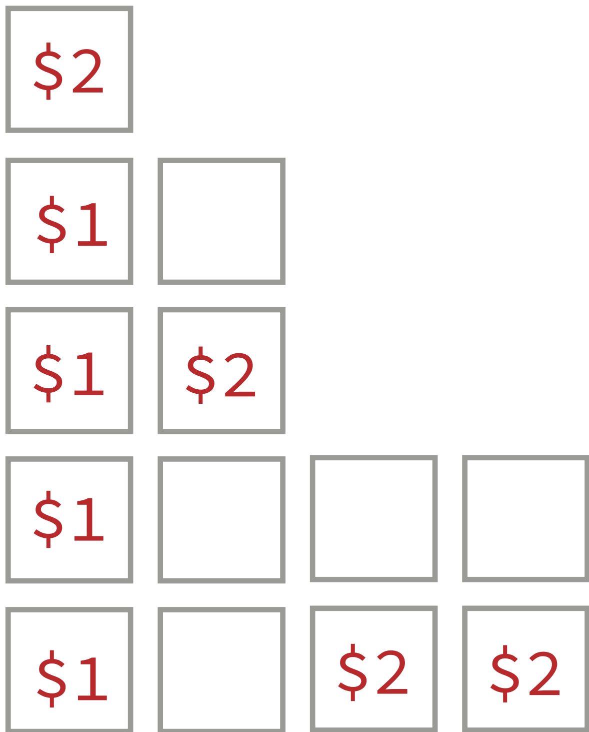
**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit

**PUSH-BACK** 2 times: use 1$ and save 2$



# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: `$1`
- Resizing the array: `$n`

**New Costs.**

- Copying a single element: `$3`
- Resizing the array: `$0`

**SECOND ATTEMPT.** Use `$3` instead of `$2`

**PUSH–BACK** use `1$` and save `2$`

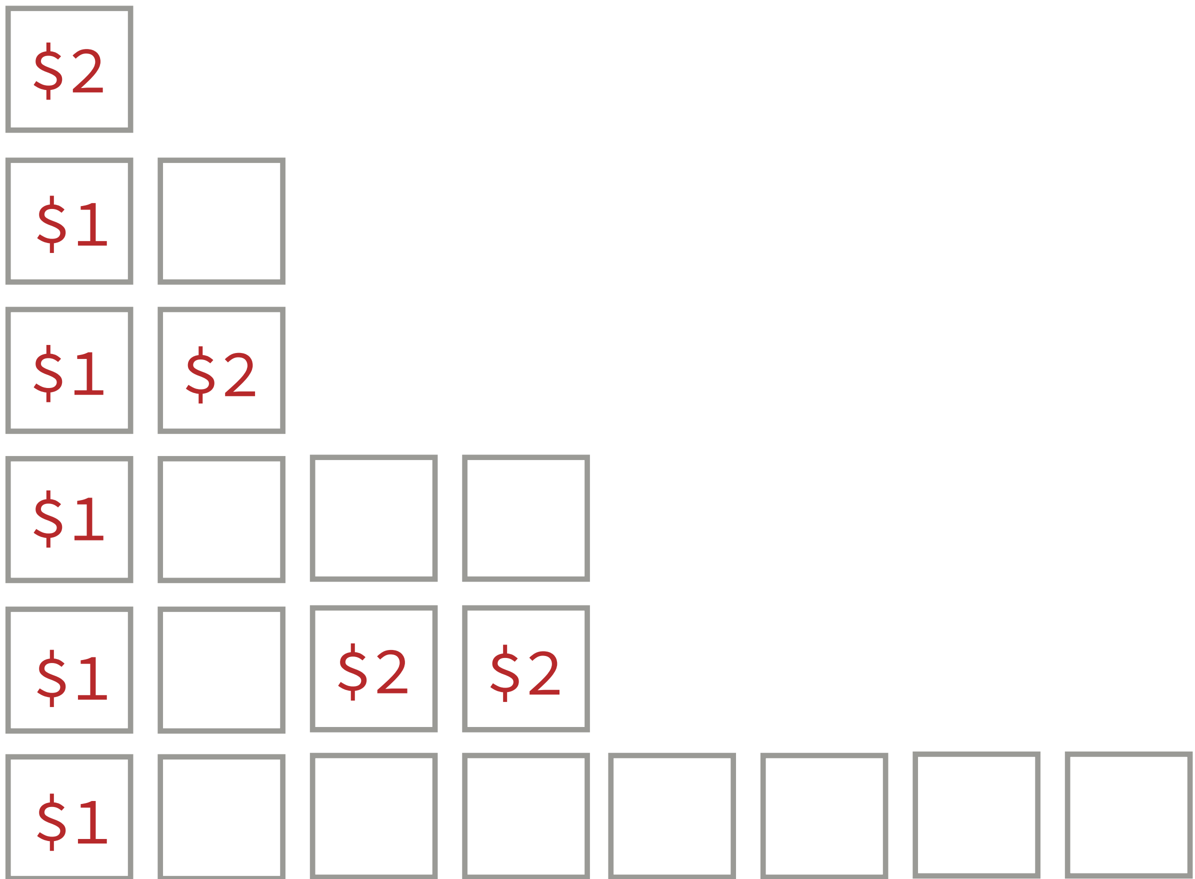
**RESIZE** use `1$` from the saved credit

**PUSH–BACK** use `1$` and save `2$`

**RESIZE** use `2$` from the saved credit

**PUSH–BACK** 2 times: use `1$` and save `2$`

**RESIZE** use `4$` from the saved credit to copy 4 elements



# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

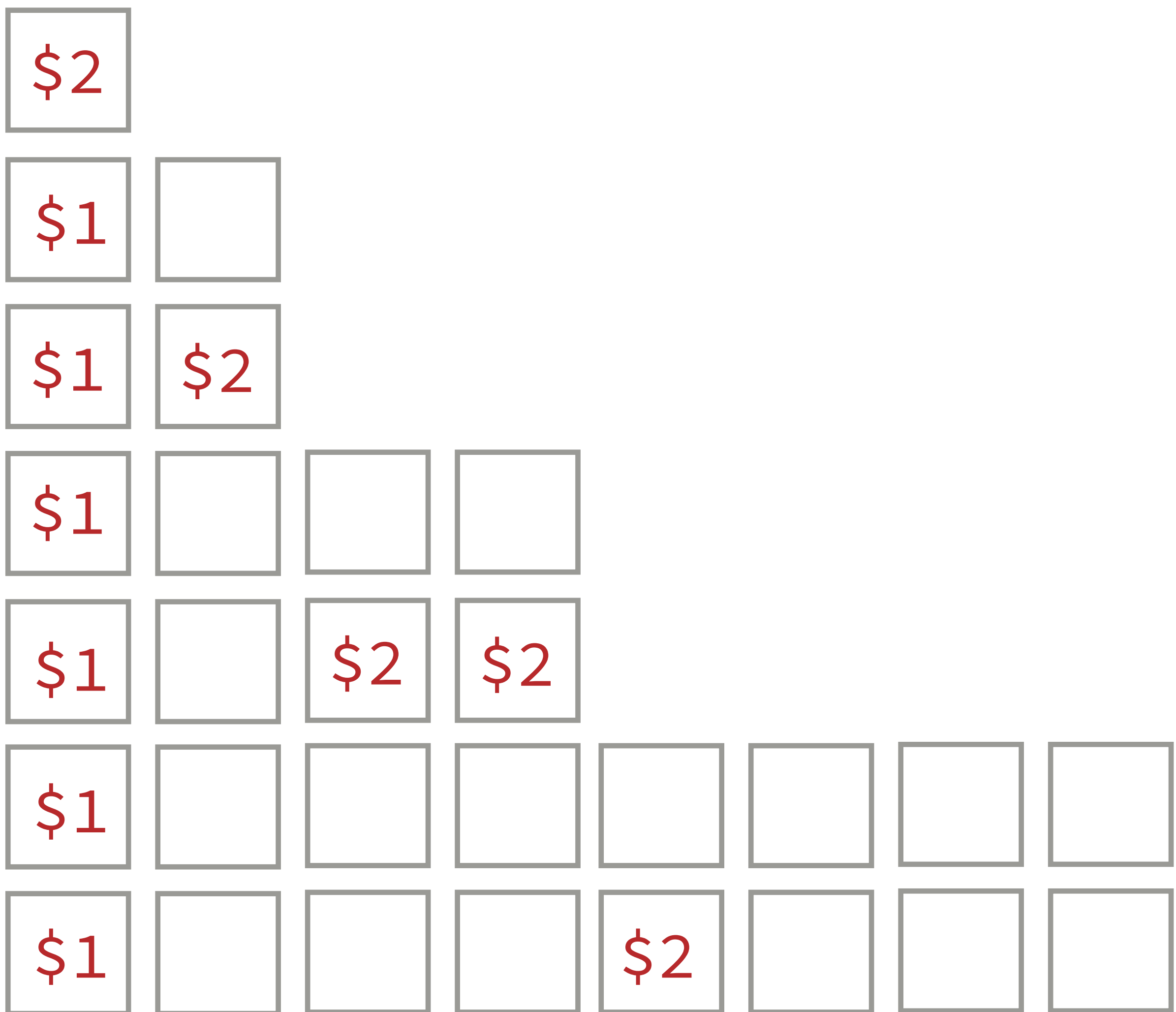
**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit

**PUSH-BACK** 2 times: use 1$ and save 2$

**RESIZE** use 4$ from the saved credit

**PUSH-BACK** use 1$ and save 2$



# Accountants Method: Resizing Array Demo

**Actual Costs.**
- Copying a single element: $1
- Resizing the array: $n

**New Costs.**
- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

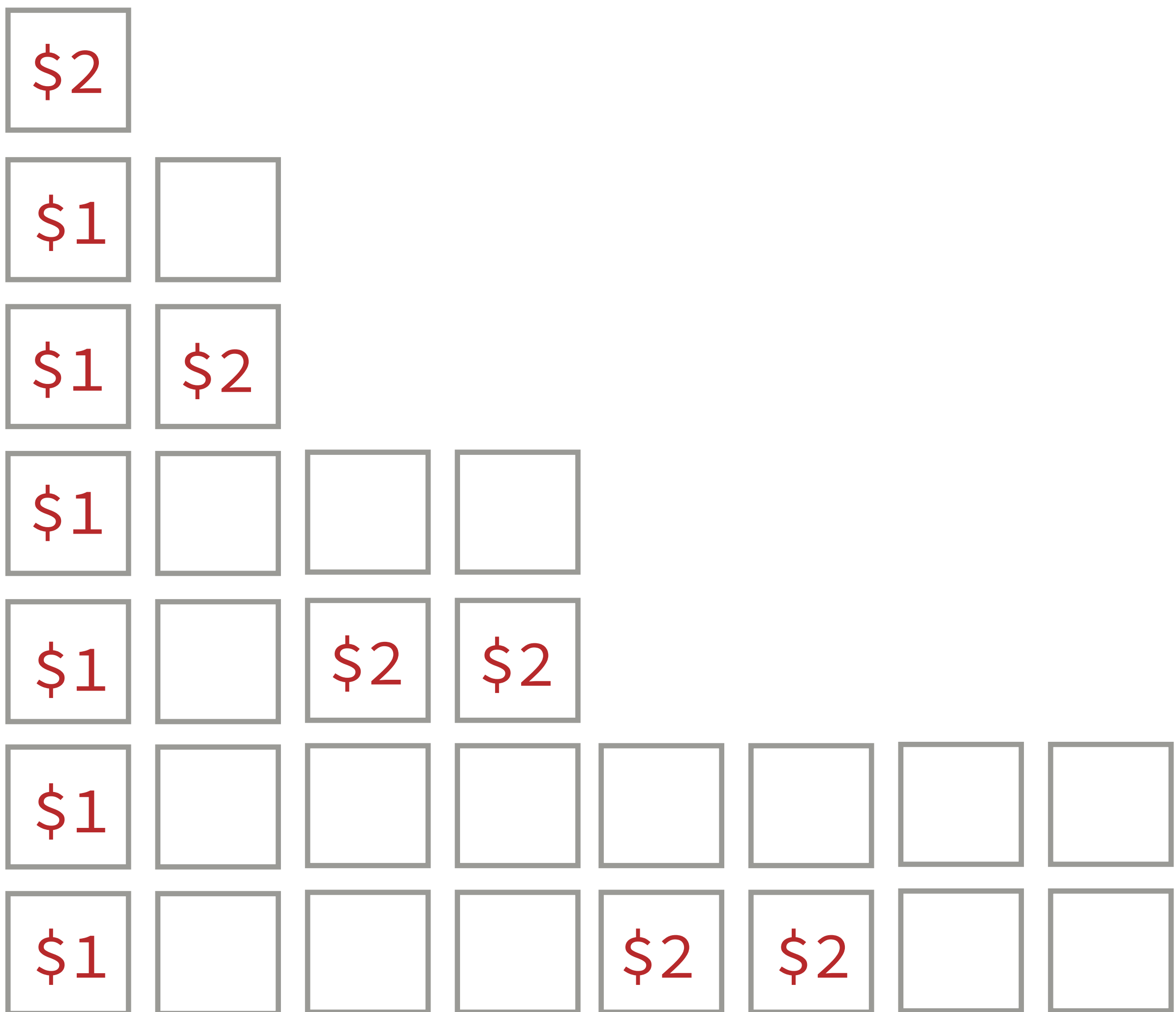
**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit

**PUSH-BACK** 2 times: use 1$ and save 2$

**RESIZE** use 4$ from the saved credit

**PUSH-BACK** 2 times: use 1$ and save 2$



# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

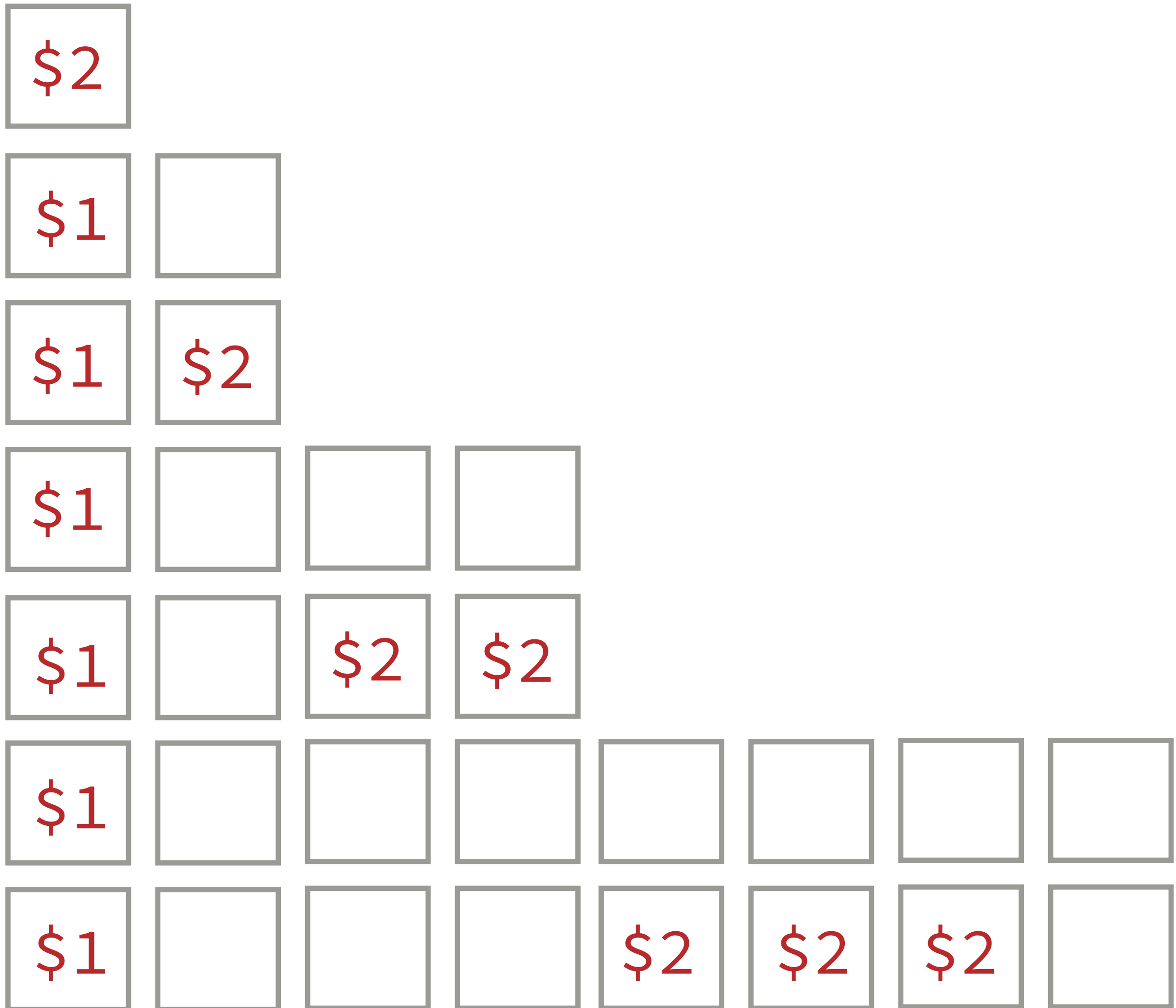
**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit

**PUSH-BACK** 2 times: use 1$ and save 2$

**RESIZE** use 4$ from the saved credit

**PUSH-BACK** 3 times: use 1$ and save 2$



# Accountants Method: Resizing Array Demo

**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**SECOND ATTEMPT.** Use $3 instead of $2

**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

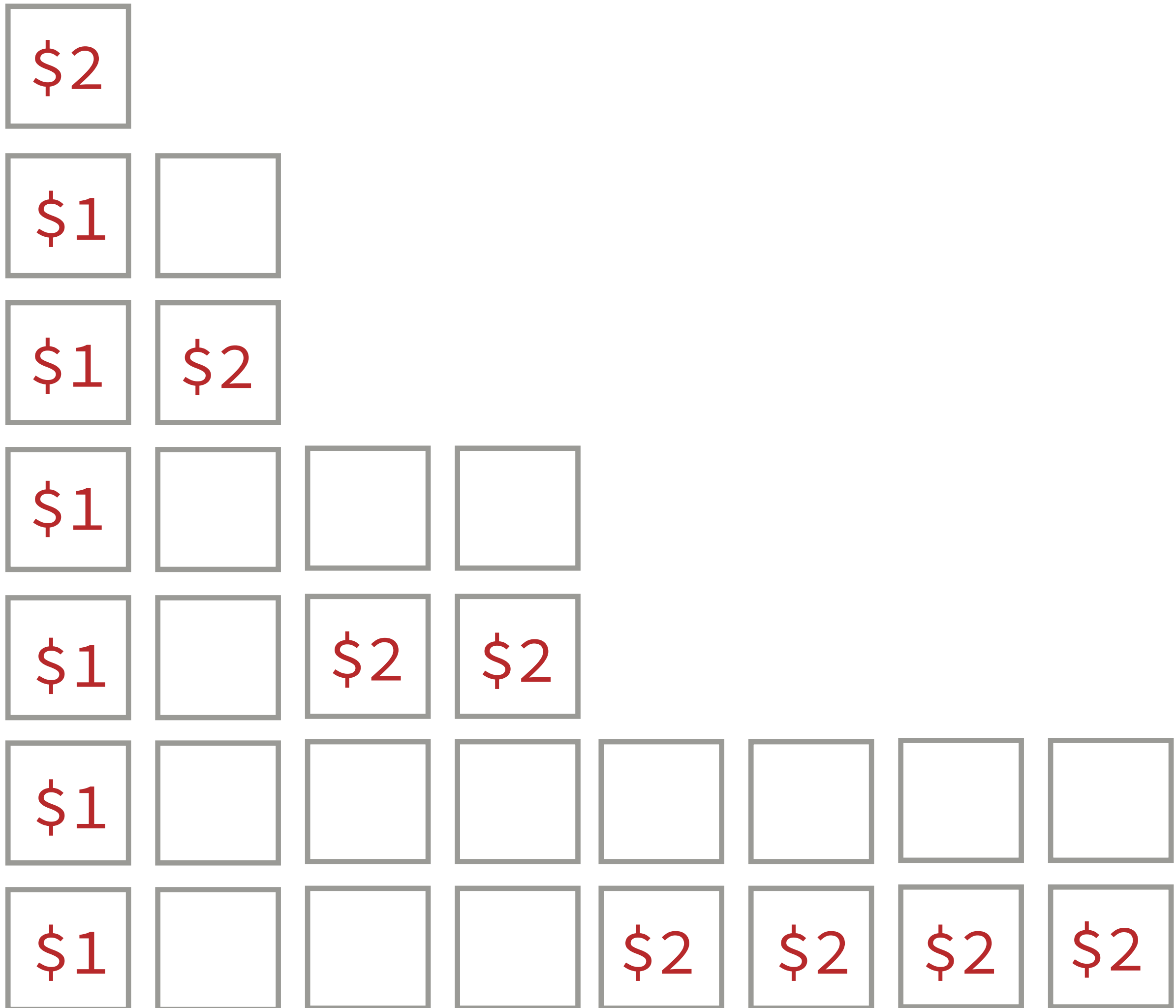
**PUSH-BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit

**PUSH-BACK** 2 times: use 1$ and save 2$

**RESIZE** use 4$ from the saved credit

**PUSH-BACK** 4 times: use 1$ and save 2$



# **Accountants Method:** Resizing Array Demo

**Actual Costs.**

- Copying a single element: `$1`
- Resizing the array: `$n`

**New Costs.**

- Copying a single element: `$3`
- Resizing the array: `$0`

**SECOND ATTEMPT.** Use `$3` instead of `$2`

**PUSH–BACK** use 1$ and save 2$

**RESIZE** use 1$ from the saved credit

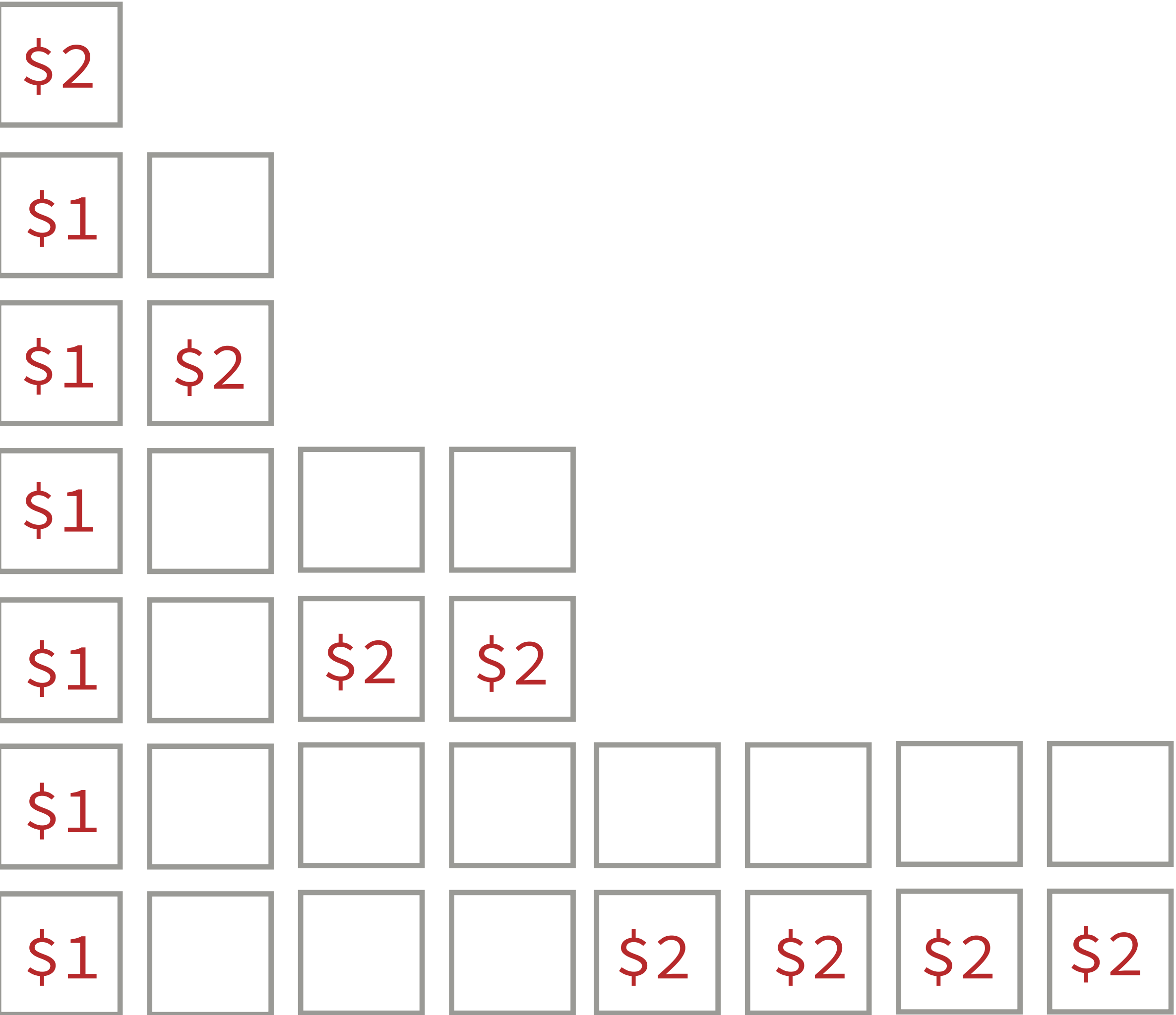
**PUSH–BACK** use 1$ and save 2$

**RESIZE** use 2$ from the saved credit

**PUSH–BACK** 2 times: use 1$ and save 2$

**RESIZE** use 4$ from the saved credit

**PUSH–BACK** 4 times: use 1$ and save 2$



There is enough to pay for a new resize!

# **Accountants Method:** Resizing Array Demo

**Actual Costs.**

- Copying a single element: `$1`
- Resizing the array: `$n`

**New Costs.**

- Copying a single element: `$3`
- Resizing the array: `$0`

Claim. Credit will always remain nonnegative.

Proof. From the examples in the previous slide, this is true for 9 **PUSH-BACK** operations. Assume that the claim is true for $k$ **PUSH-BACK** operations.

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**Note that:**

- Copying a single element without resize adds **$2** to the credit.
- Resizing produces an array of size $n$, where $\frac{n}{2}$ cells are empty.

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Hence, the claim is true for the $k+1$ **PUSH−BACK** operation.

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**Actual Costs.**

- Copying a single element: $1
- Resizing the array: $n

**New Costs.**

- Copying a single element: $3
- Resizing the array: $0

**Amortized Cost.**

- Copying a single element: O(1)
- Resizing the array: O(1)

Claim. Credit will always remain nonnega

Proof. From the examples in the previous
operations. Assume that the claim is true

**Note that:**

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- Resizing the array: `$0`

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*Assuming arrays are not resized unless they are full and that their size doubles when they resize.*

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*Assuming arrays are not resized unless they are full and that their size doubles when they resize.*

Claim. Credit will always remain nonnega

Proof. From the examples in the previous operations. Assume that the claim is true

**Interpretation.**

Any sequence of **PUSH−BACK** operations performs *at most* $\Theta(n)$ operations in total.

**Note that:**

- Copying a single element without
- Resizing produces an array of size
- Therefore, the next resize happens $2 $\times \frac{n}{2} =$ $n to the credit.
- This is enough to copy the $n$ elements in the next resize operation.

Hence, the claim is true for the $k + 1$ **PUSH−BACK** operation.

# **Accountants Method:** Incrementing a Binary Counter

**Actual Costs.**
- Flipping `0`'s to `1`'s:
- Flipping `1`'s to `0`'s: up to

**New Costs.**
- Flipping `0`'s to `1`'s:
- Flipping `1`'s to `0`'s:

| | | | | | |
|---|---|---|---|---|---|
| 0 | 0 | 0 | 0 | 0 | 0 |
| 1 | 0 | 0 | 0 | 0 | 1 |
| 2 | 0 | 0 | 0 | 1 | 0 |
| 3 | 0 | 0 | 0 | 1 | 1 |
| 4 | 0 | 0 | 1 | 0 | 0 |
| 5 | 0 | 0 | 1 | 0 | 1 |
| 6 | 0 | 0 | 1 | 1 | 0 |
| 7 | 0 | 0 | 1 | 1 | 1 |
| 8 | 0 | 1 | 0 | 0 | 0 |
| 9 | 0 | 1 | 0 | 0 | 1 |
| 10 | 0 | 1 | 0 | 1 | 0 |
| 11 | 0 | 1 | 0 | 1 | 1 |
| 12 | 0 | 1 | 1 | 0 | 0 |
| 13 | 0 | 1 | 1 | 0 | 1 |
| 14 | 0 | 1 | 1 | 1 | 0 |
| 15 | 0 | 1 | 1 | 1 | 1 |

Flipping 0's is cheap! (rows 4–5)

Flipping 1's can be expensive! (rows 7–8)

# Accountants Method: Incrementing a Binary Counter

**Actual Costs.**

- Flipping `0`'s to `1`'s: `$1`
- Flipping `1`'s to `0`'s: up to `$log(n)`

**New Costs.**

- Flipping `0`'s to `1`'s:
- Flipping `1`'s to `0`'s:

```
0   0 0 0 0 0
1   0 0 0 0 1
2   0 0 0 1 0
3   0 0 0 1 1
4   0 0 1 0 0
5   0 0 1 0 1
6   0 0 1 1 0
7   0 0 1 1 1
8   0 1 0 0 0
9   0 1 0 0 1
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**New Costs.**

- Flipping `0`'s to `1`'s: `$2`
- Flipping `1`'s to `0`'s: `$0`

**Let's Try!**

| | | | | | |
|---|---|---|---|---|---|
| 0 | 0 | 0 | 0 | 0 | 0 |
| 1 | 0 | 0 | 0 | 0 | 1 |
| 2 | 0 | 0 | 0 | 1 | 0 |
| 3 | 0 | 0 | 0 | 1 | 1 |
| 4 | 0 | 0 | 1 | 0 | 0 |
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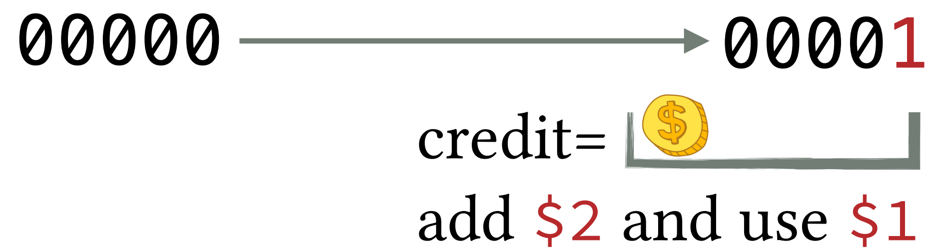

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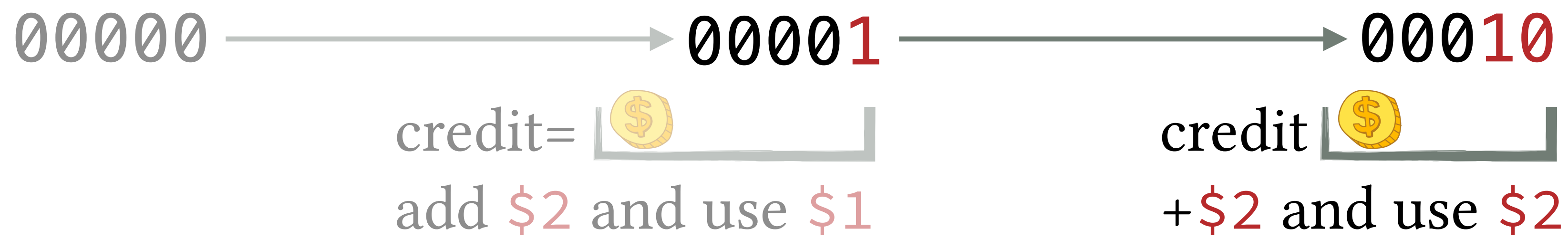

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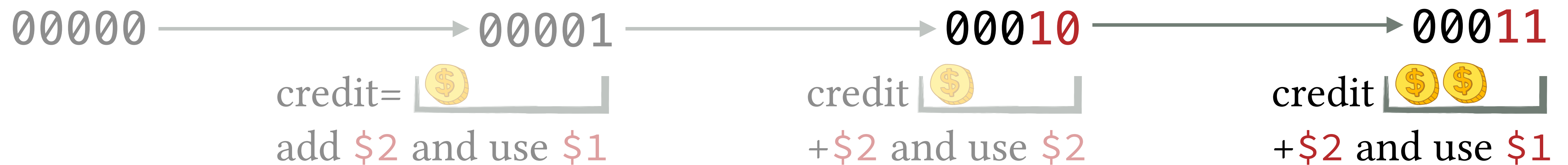

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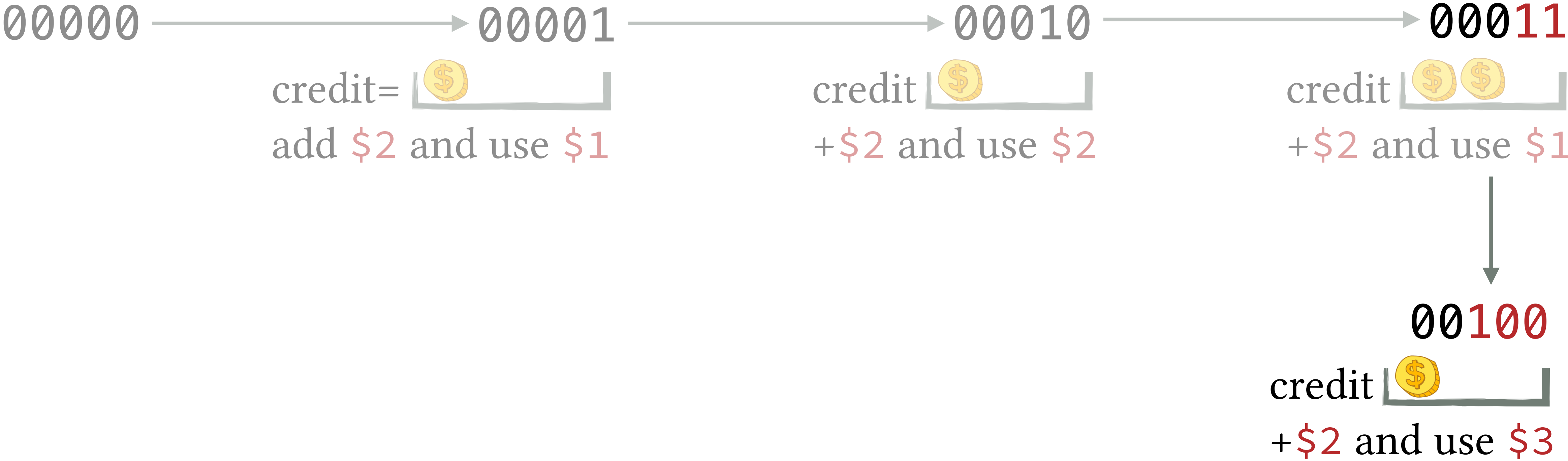

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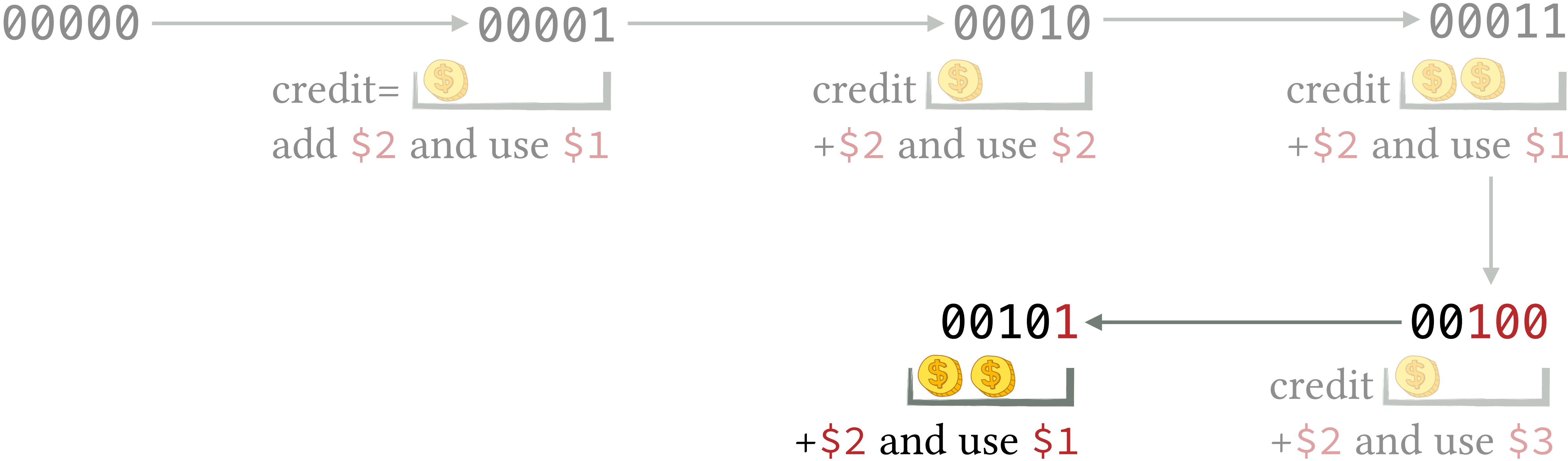

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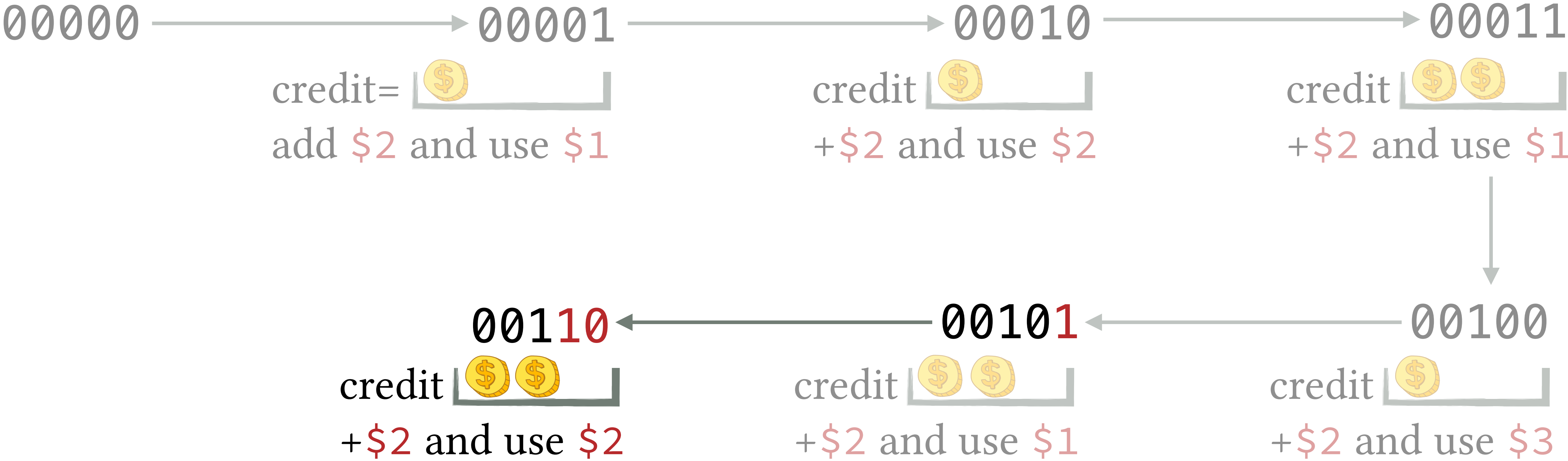

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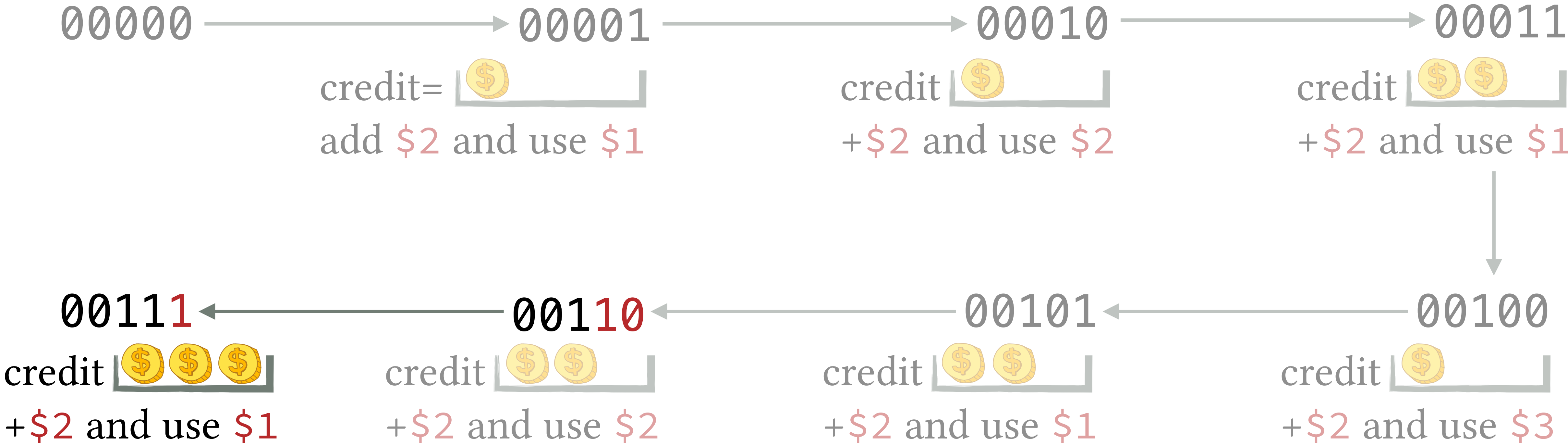

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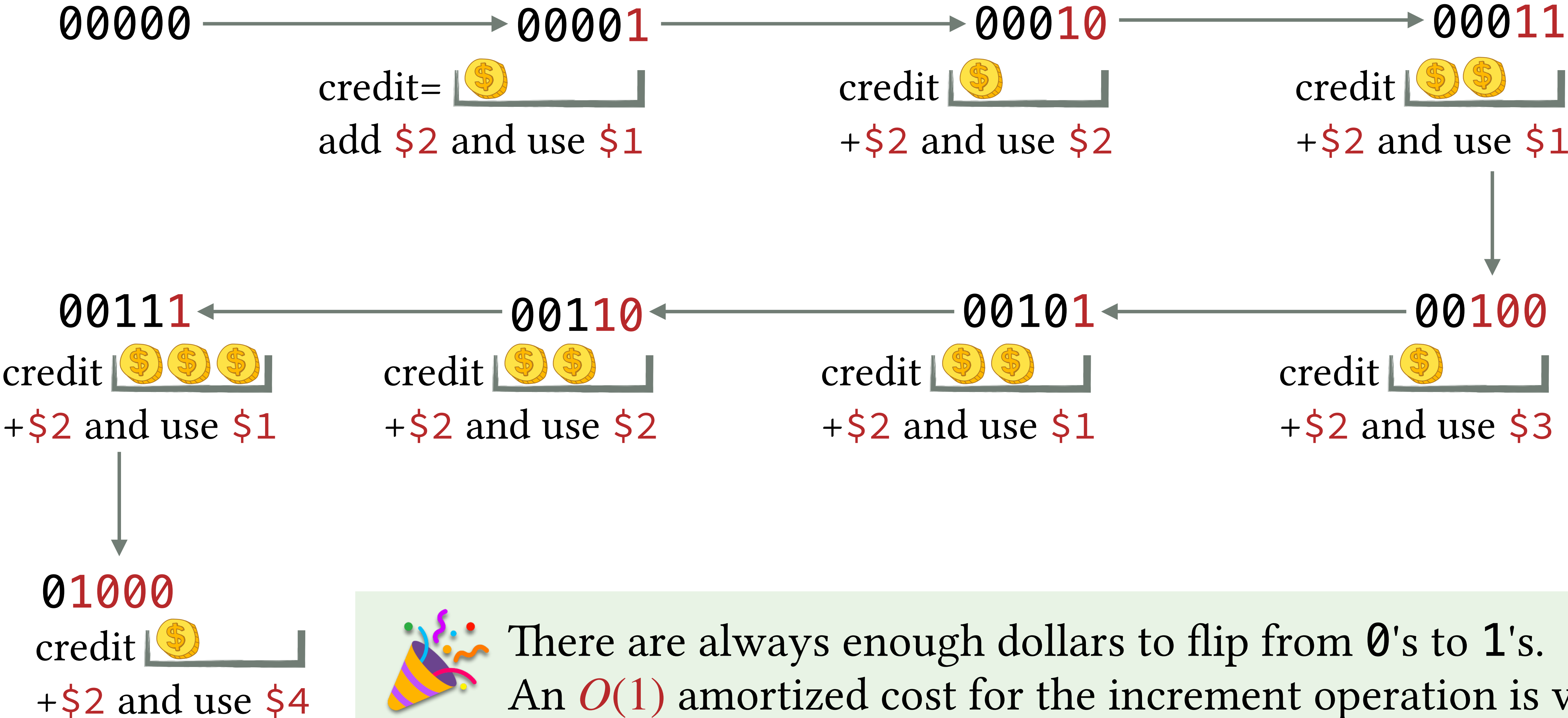


There are always enough dollars to flip from 0's to 1's.
An $O(1)$ amortized cost for the increment operation is valid!

Whenever we get a 1, we store an extra $1 to flip the 1 back to a 0 later on

# **Example:** A Queue using Two Stacks

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```
CLASS Queue

Define in  as a Stack
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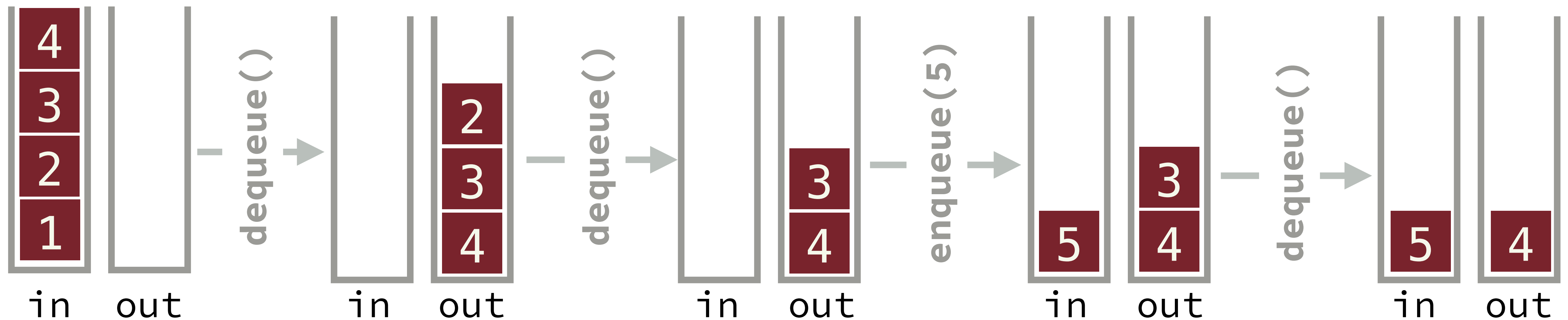
ENQUEUE(x):
  in.PUSH(x)

DEQUEUE(){
  if (out.IS-EMPTY()):
      while (not in.IS-EMPTY()):
          out.PUSH(in.POP());

  return out.POP();
```



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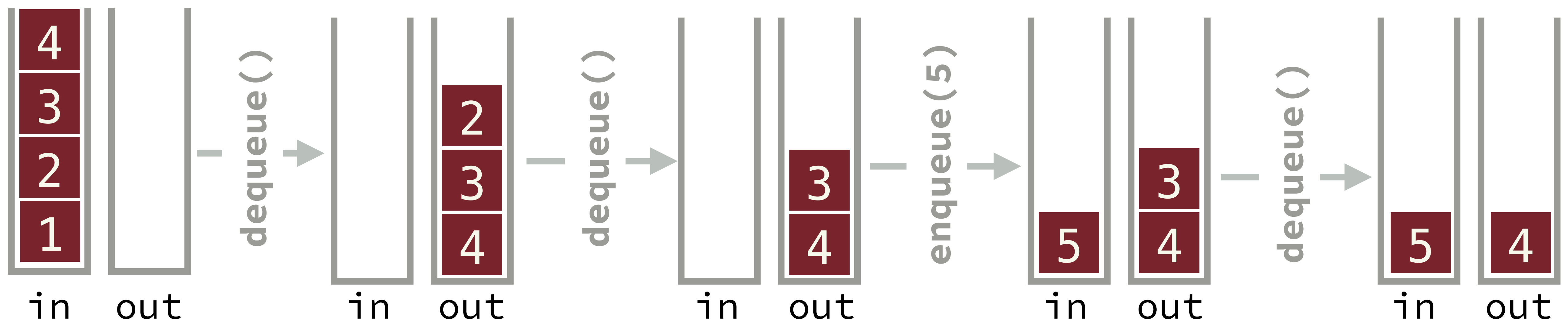
**Claim.** Any sequence of $N$ **ENQUEUE** and **DEQUEUE** operations runs in $\Theta(N)$ in the worst case.

**Exercise.** Use the Accounting Method to show that the amortized cost for each of the **ENQUEUE** and **DEQUEUE** operations is $O(1)$.



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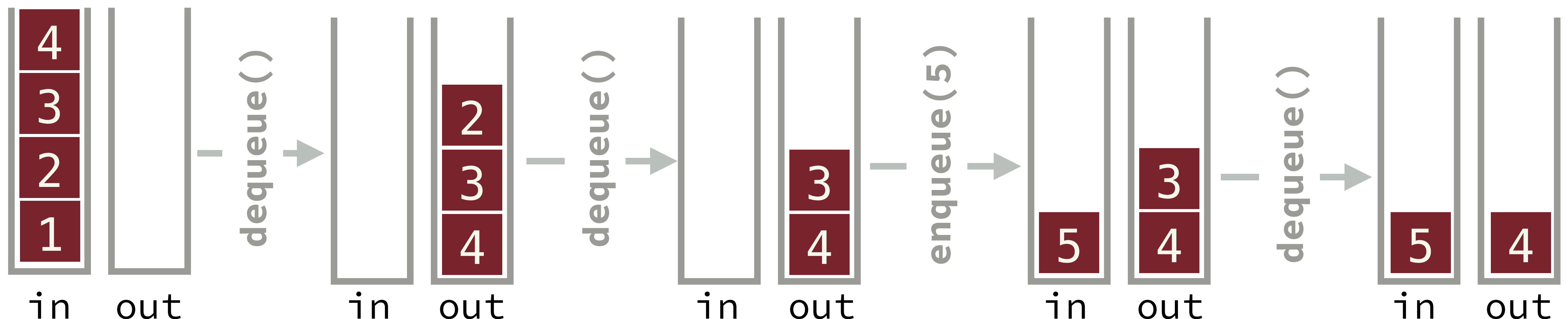
**Exercise.** Use the Accounting Method to show that the amortized cost for each of the **ENQUEUE** and **DEQUEUE** operations is $O(1)$.

**Solution.**

Pay $3 for each **ENQUEUE** operation and $0 for each **DEQUEUE** operation.

Each enqueued item will have $2 saved that can be used later for moving it to the out stack and then for popping it.



# PITFALL

Confusing average case analysis with amortized analysis.

**Average Case Analysis.** Uses probabilistic assumptions to describe how an algorithm is *expected* to behave.

**Example.** The statement "the average case for quicksort is $\Theta(n \log n)$" provides an *expectation* for the performance assuming the probability of all element permutations is the same. The algorithm is **not** guaranteed to have this performance.

**Amortized Analysis.** Does not make any probabilistic assumptions. Provides a *guarantee* for the performance of a sequence of operations.

**Example.** The statement "The **PUSH** and **POP** operations run in $O(1)$ amortized time" mean that every possible sequence of **PUSH** and **POP** operations is guaranteed to have an average running time of $O(1)$ per operation.

# The Potential Method

Definition. Given a sequence of $n$ operations, we define $\Phi(i)$ as a non-negative function that describes the potential after operation $i$, where the amortized cost of operation $i$ is:

$$\hat{c}_i = c_i + \Phi(i) - \Phi(i-1)$$

change in potential

amortized cost of operation $i$

actual cost of operation $i$

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**Claim.** If $\Phi(0) = 0$ and $\Phi(i) \geq 0$ then $\sum_{i=1}^{n} \hat{c}_i \geq \sum_{i=1}^{n} c_i$

I.e. the amortized cost for each operation is an upper bound for the actual cost!

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**Proof.** $\sum_{i=1}^{n} \hat{c}_i = \sum_{i=1}^{n} c_i + \Phi(i) - \Phi(i-1)$

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**Proof.** $\sum_{i=1}^{n} \hat{c}_i = \sum_{i=1}^{n} c_i + \Phi(i) - \Phi(i-1) = \sum_{i=1}^{n} c_i + \sum_{i=1}^{n} (\Phi(i) - \Phi(i-1))$

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$$= \sum_{i=1}^{n} c_i + \Phi(1) - \Phi(0) + \Phi(2) - \Phi(1) + \Phi(3) - \Phi(2) + \Phi(4) - \Phi(3) + \ldots + \Phi(n) - \Phi(n-1)$$

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**Proof.**

$$\sum_{i=1}^{n} \hat{c}_i = \sum_{i=1}^{n} c_i + \Phi(i) - \Phi(i-1) = \sum_{i=1}^{n} c_i + \sum_{i=1}^{n} (\Phi(i) - \Phi(i-1))$$

$$= \sum_{i=1}^{n} c_i + \Phi(1) - \Phi(0) + \Phi(2) - \Phi(1) + \Phi(3) - \Phi(2) + \Phi(4) - \Phi(3) + \ldots + \Phi(n) - \Phi(n-1)$$

$$= \sum_{i=1}^{n} c_i - \Phi(0) + \Phi(n)$$

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amortized cost of operation $i$

actual cost of operation $i$

**Claim.** If $\Phi(0) = 0$ and $\Phi(i) \geq 0$ then $\sum_{i=1}^{n} \hat{c}_i \geq \sum_{i=1}^{n} c_i$

I.e. the amortized cost for each operation is an upper bound for the actual cost!

**Proof.**

$$\sum_{i=1}^{n} \hat{c}_i = \sum_{i=1}^{n} c_i + \Phi(i) - \Phi(i-1) = \sum_{i=1}^{n} c_i + \sum_{i=1}^{n} (\Phi(i) - \Phi(i-1))$$

$$= \sum_{i=1}^{n} c_i + \Phi(1) - \Phi(0) + \Phi(2) - \Phi(1) + \Phi(3) - \Phi(2) + \Phi(4) - \Phi(3) + \ldots + \Phi(n) - \Phi(n-1)$$

$$= \sum_{i=1}^{n} c_i - \Phi(0) + \Phi(n) \geq \sum_{i=1}^{n} c_i$$

# The Potential Method

Definition. Given a sequence of $n$ operations, we define $\Phi(i)$ as a non-negative function that describes the potential after operation $i$, where the amortized cost of operation $i$ is:

$$\hat{c}_i = c_i + \Phi(i) - \Phi(i-1)$$

change in potential

amortized cost of operation $i$

actual cost of operation $i$

**Claim.** If $\Phi(0) = 0$ and $\Phi(i) \geq 0$ then $\sum_{i=1}^{n} \hat{c}_i \geq \sum_{i=1}^{n} c_i$

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$$\sum_{i=1}^{n} \hat{c}_i = \sum_{i=1}^{n} c_i + \Phi(i) - \Phi(i-1) = \sum_{i=1}^{n} c_i + \sum_{i=1}^{n} (\Phi(i) - \Phi(i-1))$$

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$$= \sum_{i=1}^{n} c_i - \Phi(0) + \Phi(n) \geq \sum_{i=1}^{n} c_i$$

🤔 **How do we choose the** Potential **function?**

# Example # 1: Resizing Arrays

Bad Choice of $\Phi(i)$. Let $\Phi(i)$ be equal to the number of elements in the array after applying operation operation $i$.

- $\Phi(0) = 0$ because initially the array has 0 elements
- $\Phi(i) = i$ is always non-negative

amortized cost = $O(1)$

Case # 1 - no resize: $\hat{c}_i = 1 + \Phi(i) - \Phi(i-1) = 1 + i - (i-1) = 2$

actual cost

number of elements now

number of elements before

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Case # 2 - with resize: $\hat{c}_i = i + \Phi(i) - \Phi(i-1) = i + i - (i-1) = i + 1$

actual cost

number of elements now

number of elements before

amortized cost = $O(1)$

amortized cost = $O(n)$

This upper bound is not interesting! It is the same as the actual worst-case running time we already know!

WHO CARES?

# Example # 1: Resizing Arrays

Good Choice of $\Phi(i)$. Let $N_i =$ the capacity of the array after operation $i$, and let

$$\text{let } \Phi(i) = 2(i - \frac{N_i}{2}) = 2i - N_i$$

Double the number of elements in the second half of the array (after operation $i$)

Hence:

- $\Phi(0) = 0$ because i = 0 and $N_i = 0$
- $\Phi(i) = 2i - N \geq 0$ because the array is *never* less than half full

Case # 1 - no resize: $\hat{c}_i = 1 + \Phi(i) - \Phi(i-1) = 1 + (2i - N_i) - (2(i-1) - N_{i-1})$

$$= 1 + 2i - N_i - 2i + 2 + N_{i-1} = 3$$

actual cost

capacity before and after is the same

# Example # 1: Resizing Arrays

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Case # 2 - with resize: $\hat{c}_i = i + \Phi(i) - \Phi(i-1) = i + (2i - N_i) - (2(i-1) - \frac{N_i}{2})$

actual cost (points to $i$)

capacity before is half the current capacity (points to $\frac{N_i}{2}$)

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Case # 2 - with resize: $\hat{c}_i = i + \Phi(i) - \Phi(i-1) = i + (2i - N_i) - (2(i-1) - \frac{N_i}{2})$

$$= i + 2i - N_i - 2i + 2 + \frac{N_i}{2}$$

$$= \frac{N_i}{2} + 1 - N_i + 2 + \frac{N_i}{2}$$

current number of elements = half the current capacity + 1

# Example # 1: Resizing Arrays

Good Choice of $\Phi(i)$. Let $N_i =$ the capacity of the array after operation $i$, and let

$$\text{let } \Phi(i) = 2(i - \frac{N_i}{2}) = 2i - N_i$$

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Case # 2 - with resize: $\hat{c}_i = i + \Phi(i) - \Phi(i-1) = i + (2i - N_i) - (2(i-1) - \frac{N_i}{2})$

$$= i + 2i - N_i - 2i + 2 + \frac{N_i}{2}$$

$$= \frac{N_i}{2} + 1 - N_i + 2 + \frac{N_i}{2}$$

$$= 3$$

← amortized cost = $O(1)$

This upper bound is interesting!
This is much lower than worst case running time we know, which is O($n$)

YES ! !

# Example # 2: Counting Bit-Flips

Potential Function. Let $\Phi(i) = N_i$, where $N_i$ *is* the number of 1s after operation *i.*

Hence:

- $\Phi(0) = 0$ because the number of 1s in 000000… is 0
- $\Phi(i) = N_i \geq 0$ because the number of bits is never negative

amortized cost = $O(1)$

Case # 1 - Flip $0 \rightarrow 1$: $\hat{c}_i = 1 + \Phi(i) - \Phi(i-1) = 1 + N_i - (N_i - 1) = 2$

actual cost

number of 1s now

number of 1s before

# Example # 2: Counting Bit-Flips

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Case # 1 - Flip $0 \rightarrow 1$: $\hat{c}_i = 1 + \Phi(i) - \Phi(i-1) = 1 + N_i - (N_i - 1) = 2$

Case # 2 - Flip $1 \rightarrow 0$: $\hat{c}_i = N_{i-1} + 1 + \Phi(i) - \Phi(i-1) = N_{i-1} + 1 + 1 - N_{i-1}$

in the worst case, all the 1s are flipped to 0s and one 0 is flipped to 1

in the worst case, only one 1 remains

# Example # 2: Counting Bit-Flips

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Hence:

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amortized cost = $O(1)$

Case # 1 - Flip 0 → 1: $\hat{c}_i = 1 + \Phi(i) - \Phi(i-1) = 1 + N_i - (N_i - 1) = 2$

Case # 2 - Flip 1 → 0: $\hat{c}_i = N_{i-1} + 1 + \Phi(i) - \Phi(i-1) = N_{i-1} + 1 + 1 - N_{i-1}$

$= 2$ ← amortized cost = $O(1)$

# Example # 3: A Queue using Two Stacks

Potential Function. Let $\Phi(i) = 2a_i + b_i$ , where:

- $a_i$ = number of elements in the `in` stack after operation $i$
- $b_i$ = the number of elements in the `out` stack after operation $i$

Hence:

- $\Phi(0) = 0$ — Initially, no elements in any of the stacks
- $\Phi(i) = 2a_i + b_i \geq 0$ — The number of elements is never negative

Case # 1 - Enqueue: $\hat{c}_i = 1 + \Phi(i) - \Phi(i-1) = 1 + (2a_i + b_i) - (2(a_i - 1) + b_i)$

actual cost

the `in` stack increases by 1

the `out` stack does not change

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$$= 1 + 2a_i + b_i - 2a_i + 2 - b_i$$

$$= 3$$

amortized cost = $O(1)$

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$$= 3$$

amortized cost = $O(1)$

Case # 2 - Dequeue:
assuming **out** is empty

$$\hat{c}_i = a_{i-1} + 1 + \Phi(i) - \Phi(i-1)$$

move $a_{i-1}$ elements from **in** to **out** + pop 1 element

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$$= 3$$

amortized cost = $O(1)$

Case # 2 - Dequeue:
assuming **out** is empty

$$\hat{c}_i = a_{i-1} + 1 + \Phi(i) - \Phi(i-1)$$

$$= a_{i-1} + 1 + (2 \times 0 + b_i) - (2a_{i-1} + 0)$$

**in** becomes empty

**out** was empty

# Example # 3: A Queue using Two Stacks

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$$= a_{i-1} + 1 + b_i - 2a_{i-1} =$$

# Example # 3: A Queue using Two Stacks

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$$= 1 + 2a_i + b_i - 2a_i + 2 - b_i$$

$$= 3$$

amortized cost = $O(1)$

Case # 2 - Dequeue:
assuming **out** is empty

$$\hat{c}_i = a_{i-1} + 1 + \Phi(i) - \Phi(i-1)$$

$$= a_{i-1} + 1 + (2 \times 0 + b_i) - (2a_{i-1} + 0)$$

$$= a_{i-1} + 1 + b_i - 2a_{i-1} = 1 + b_i - a_{i-1}$$

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$$= 1 + 2a_i + b_i - 2a_i + 2 - b_i$$
$$= 3$$

amortized cost = $O(1)$

Case # 2 - Dequeue:
assuming **out** is empty

$$\hat{c}_i = a_{i-1} + 1 + \Phi(i) - \Phi(i-1)$$
$$= a_{i-1} + 1 + (2 \times 0 + b_i) - (2a_{i-1} + 0)$$
$$= a_{i-1} + 1 + b_i - 2a_{i-1} = 1 + b_i - a_{i-1} = 0$$

amortized cost = $O(1)$

# of elements in **out** now is 1 less than the # of elements in **in** before