

CP312

Algorithm Design and Analysis I

LECTURE 2: ANALYZING ALGORITHMS

Recall: The Sorting Problem

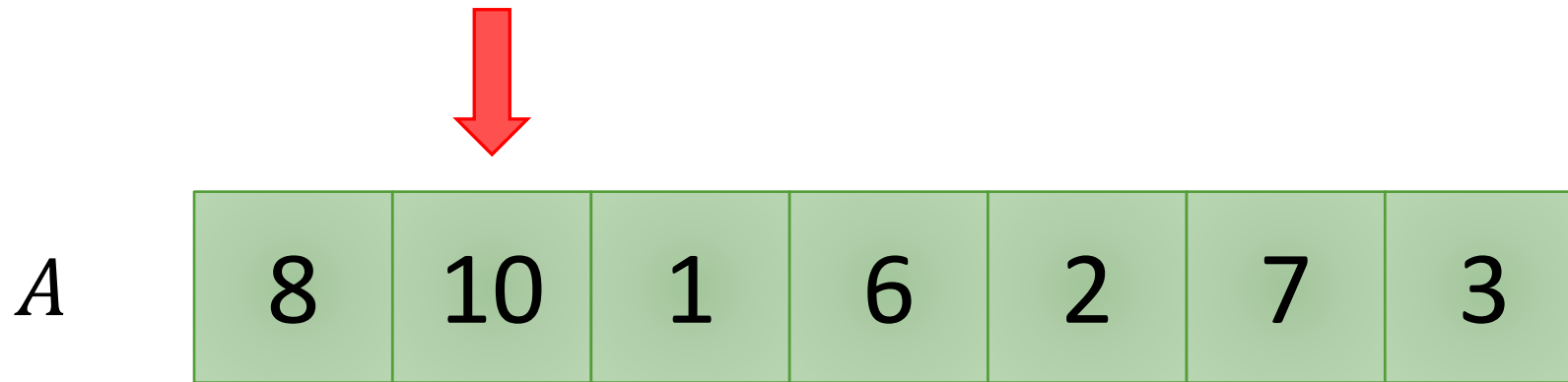
- **Problem:** Sort a sequence of numbers in non-decreasing order
- **Input:** A sequence of numbers $\pi = (a_1, \dots, a_n)$
- **Output:** A permutation $\pi' = (a'_1, \dots, a'_n)$ of π such that
$$a'_1 \leq a'_2 \leq \dots \leq a'_n$$
- An algorithm for the sorting problem is a sequence of computational steps with the above input/output specifications.
- Ex: $(8, 10, 1, 6, 2, 7, 3) \Rightarrow (1, 2, 3, 6, 7, 8, 10)$

Insertion Sort

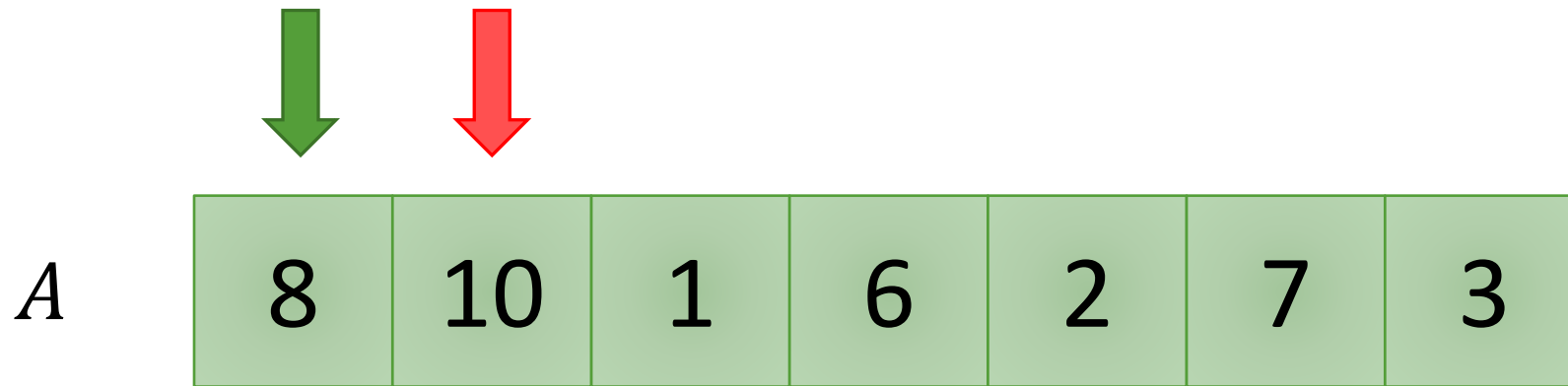
A

8	10	1	6	2	7	3
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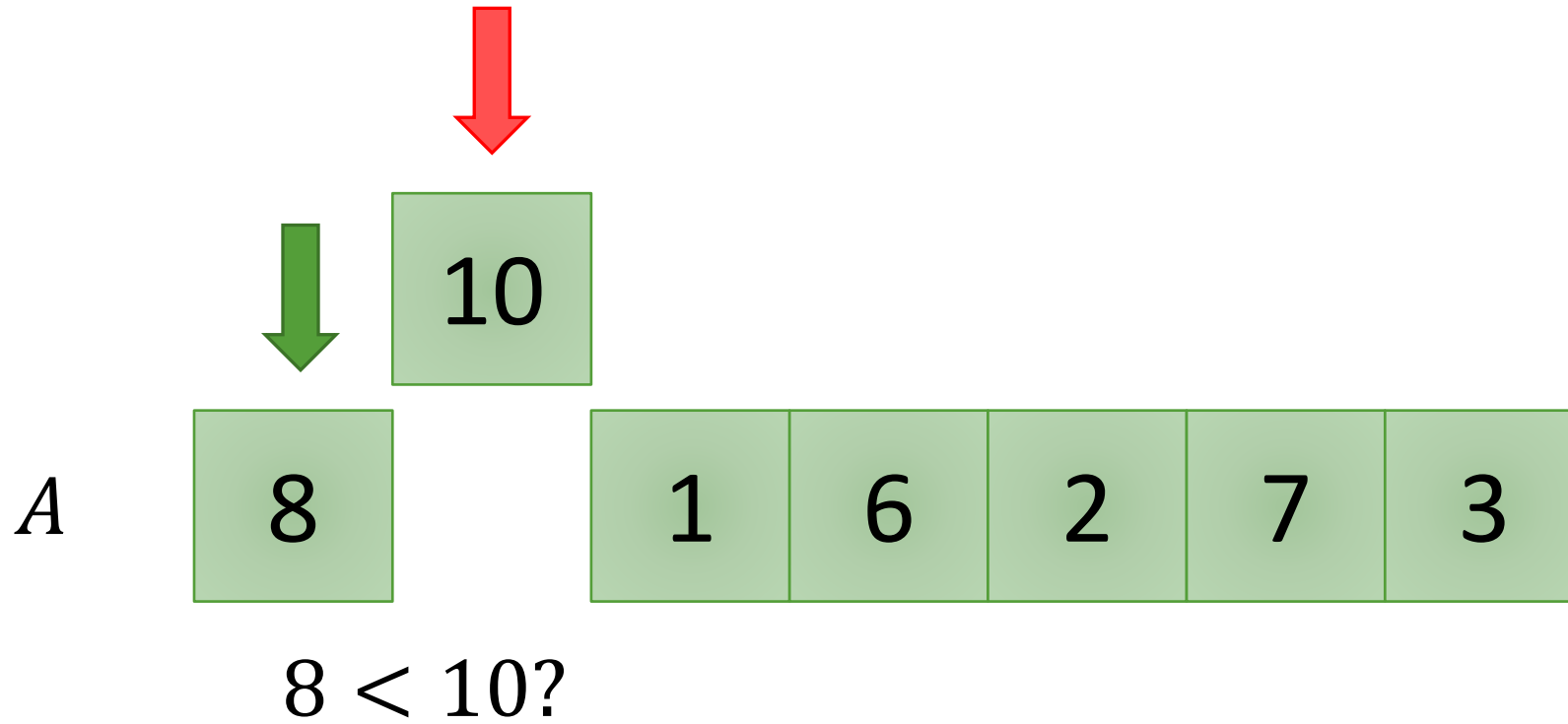
Insertion Sort



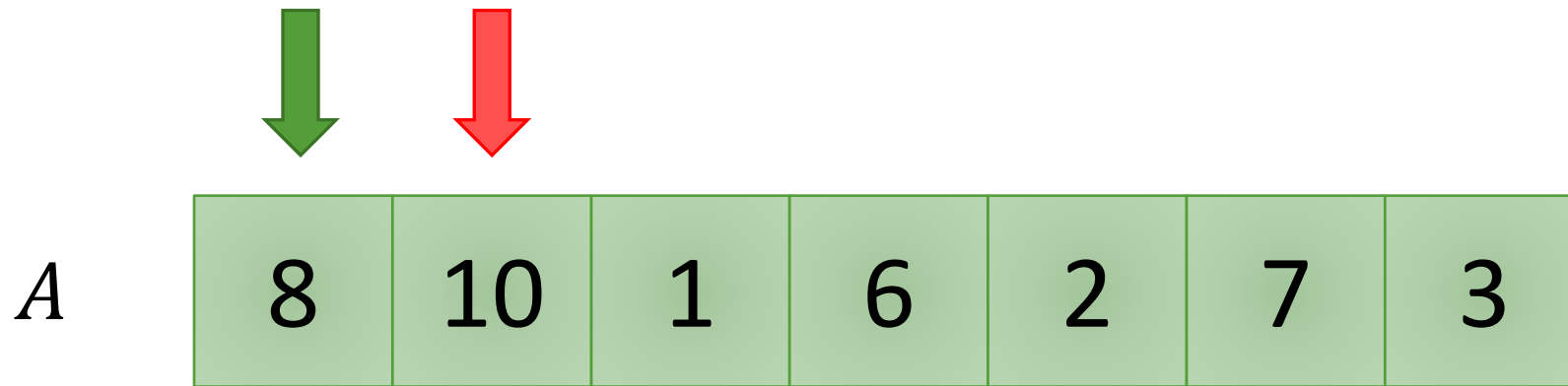
Insertion Sort



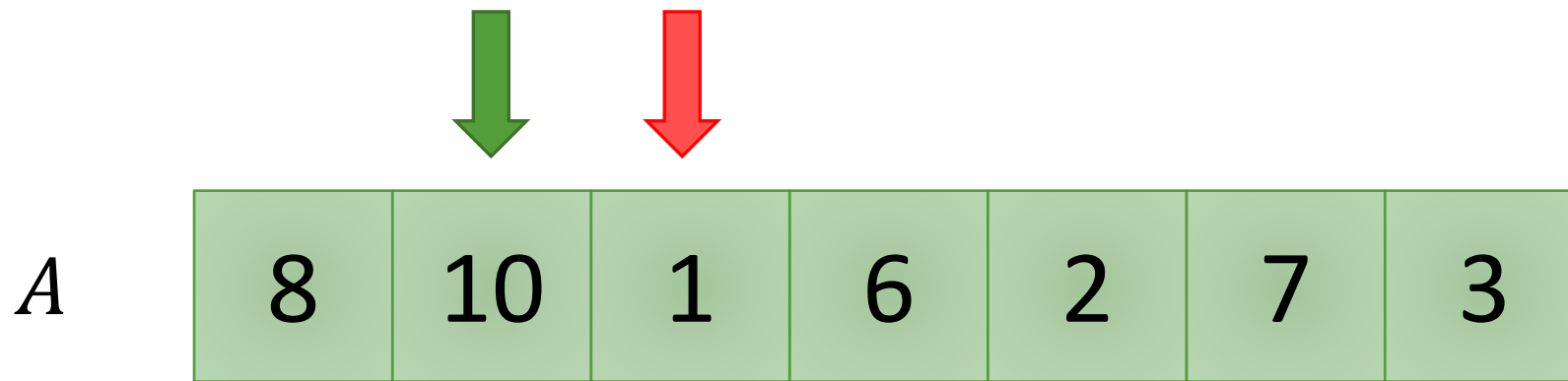
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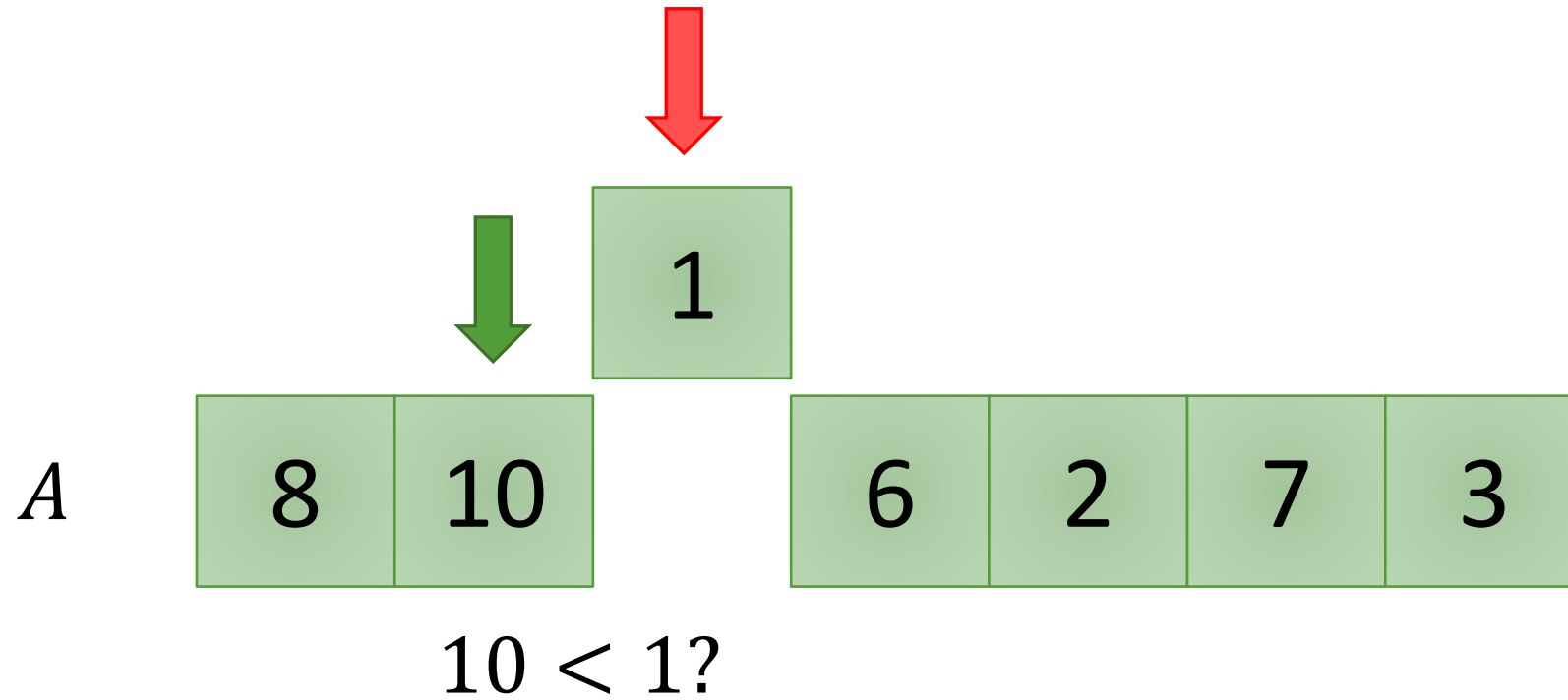
Insertion Sort



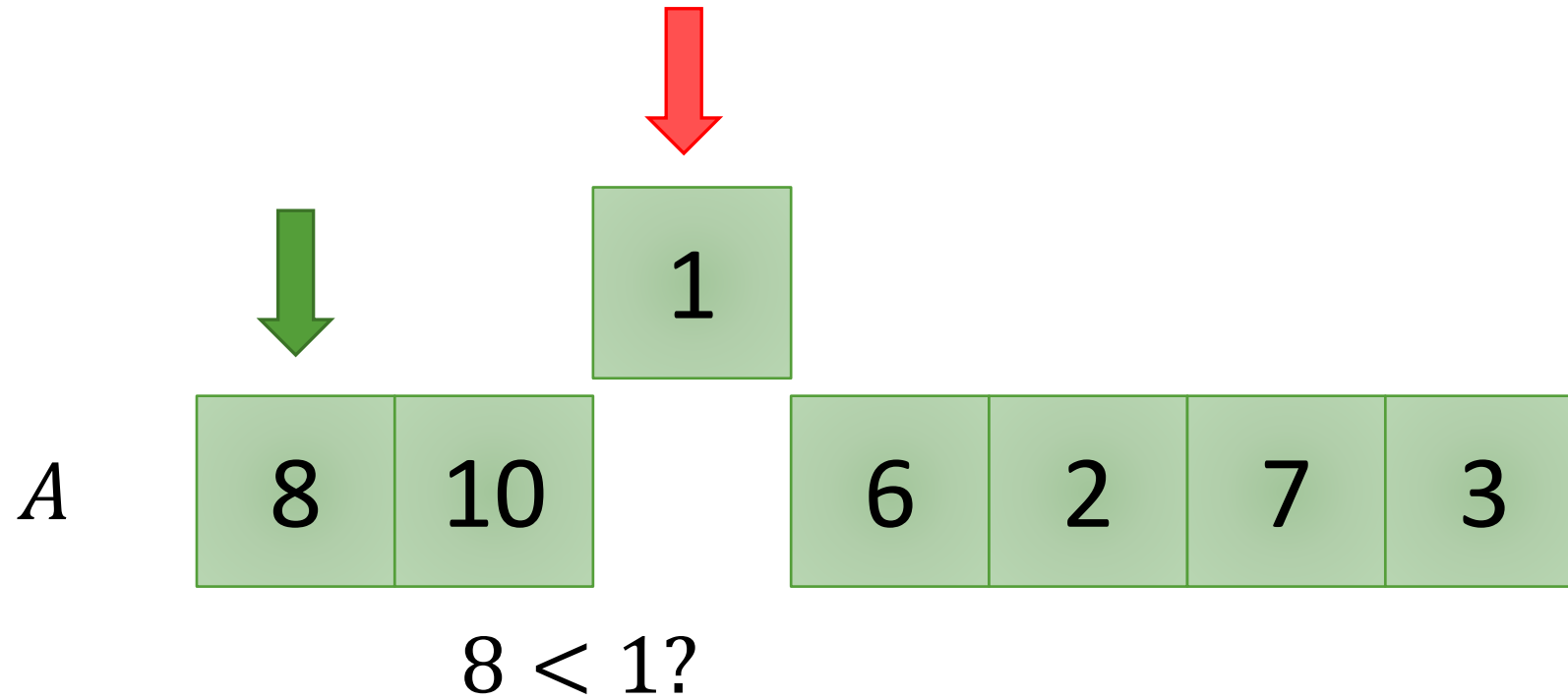
Insertion Sort



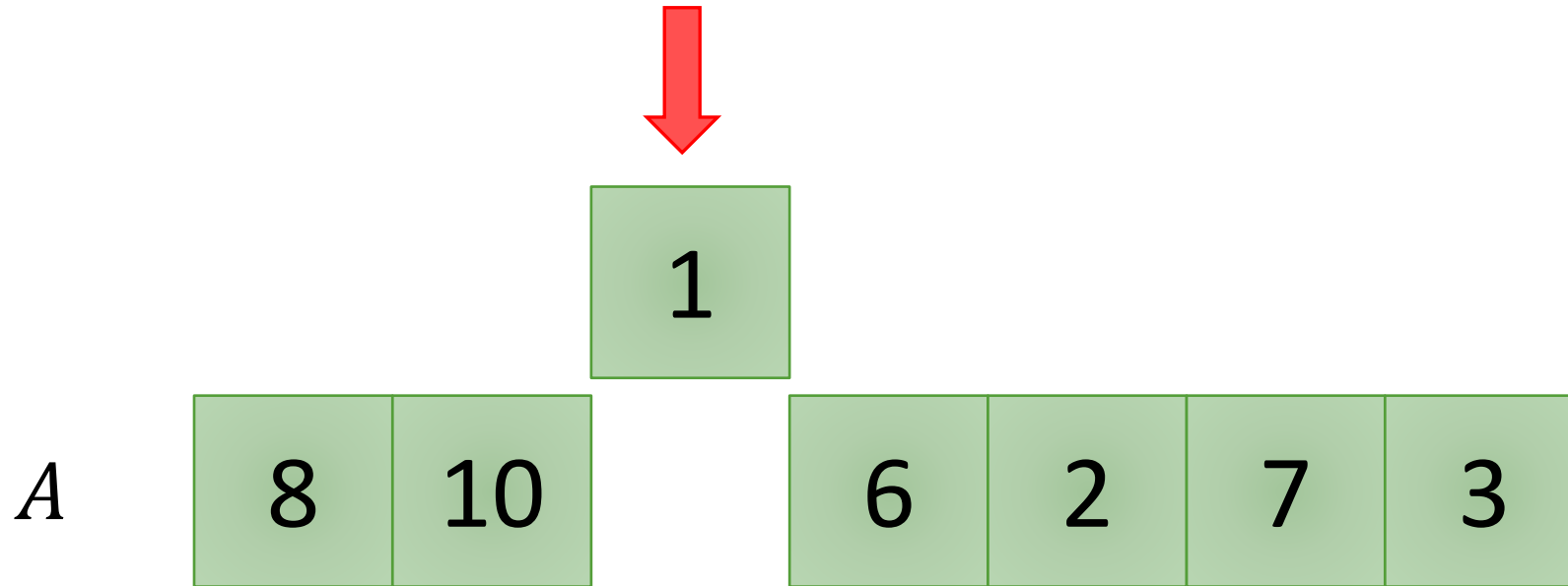
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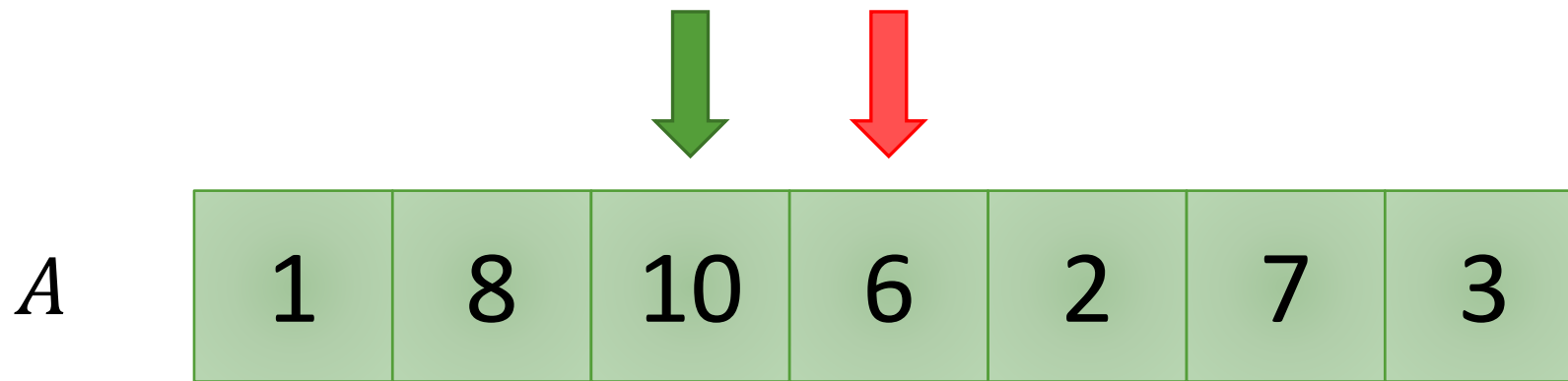
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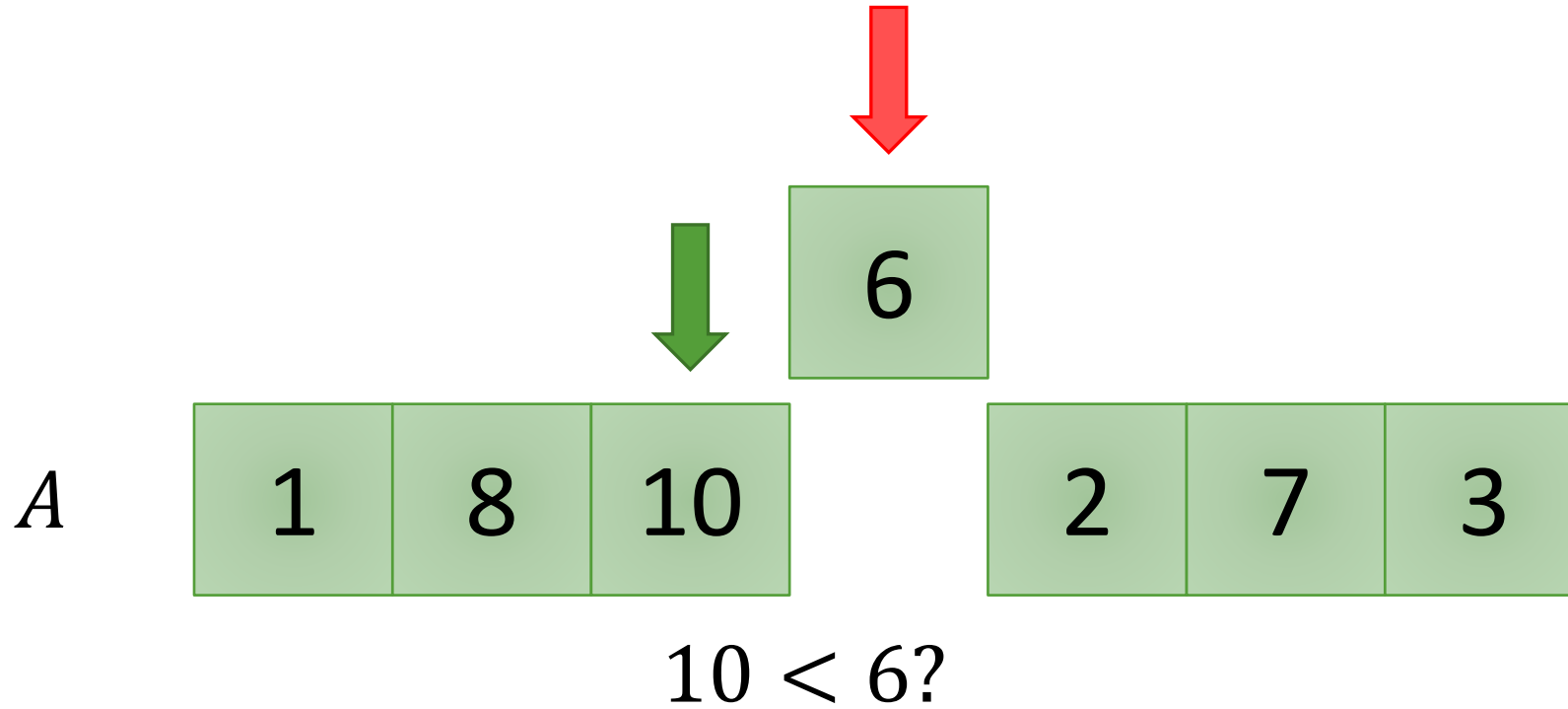
Insertion Sort



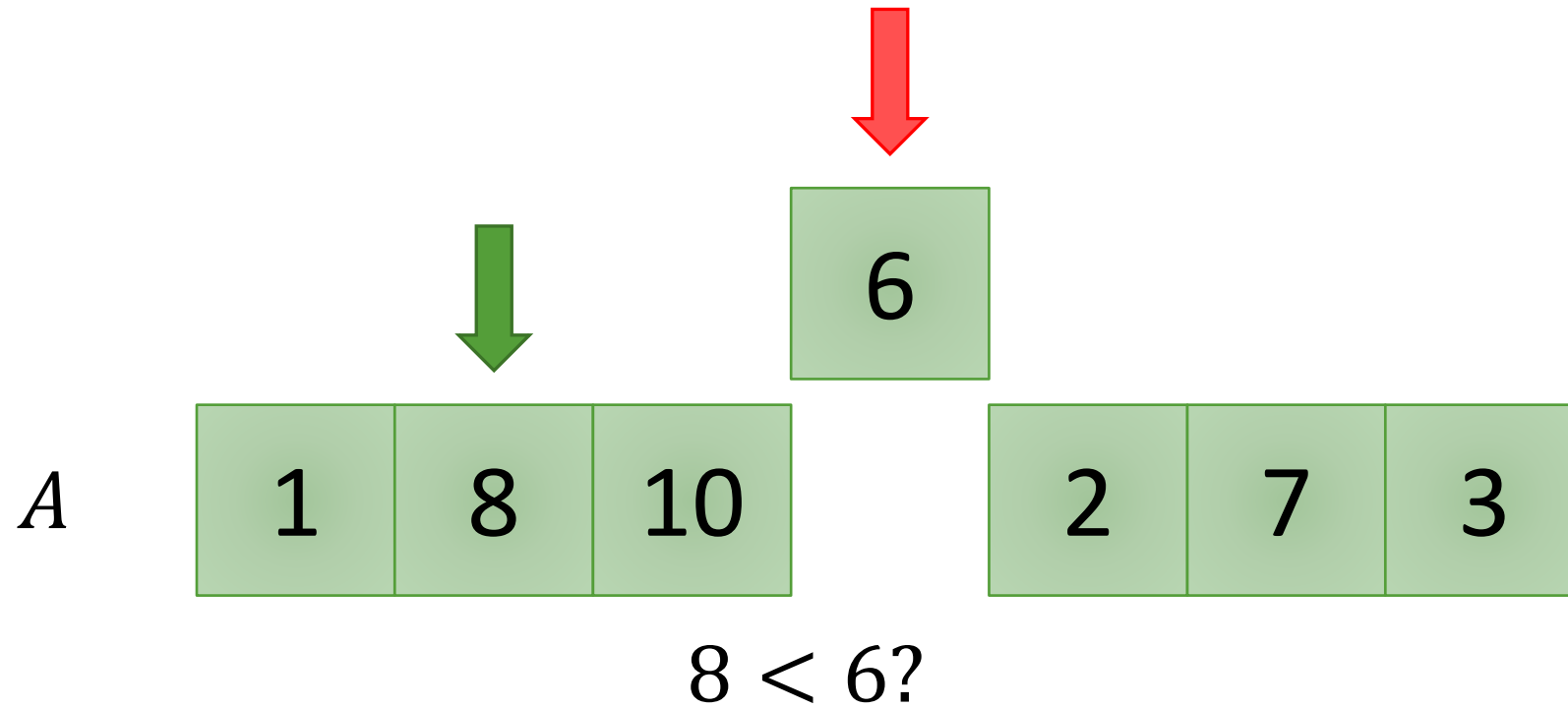
Insertion Sort



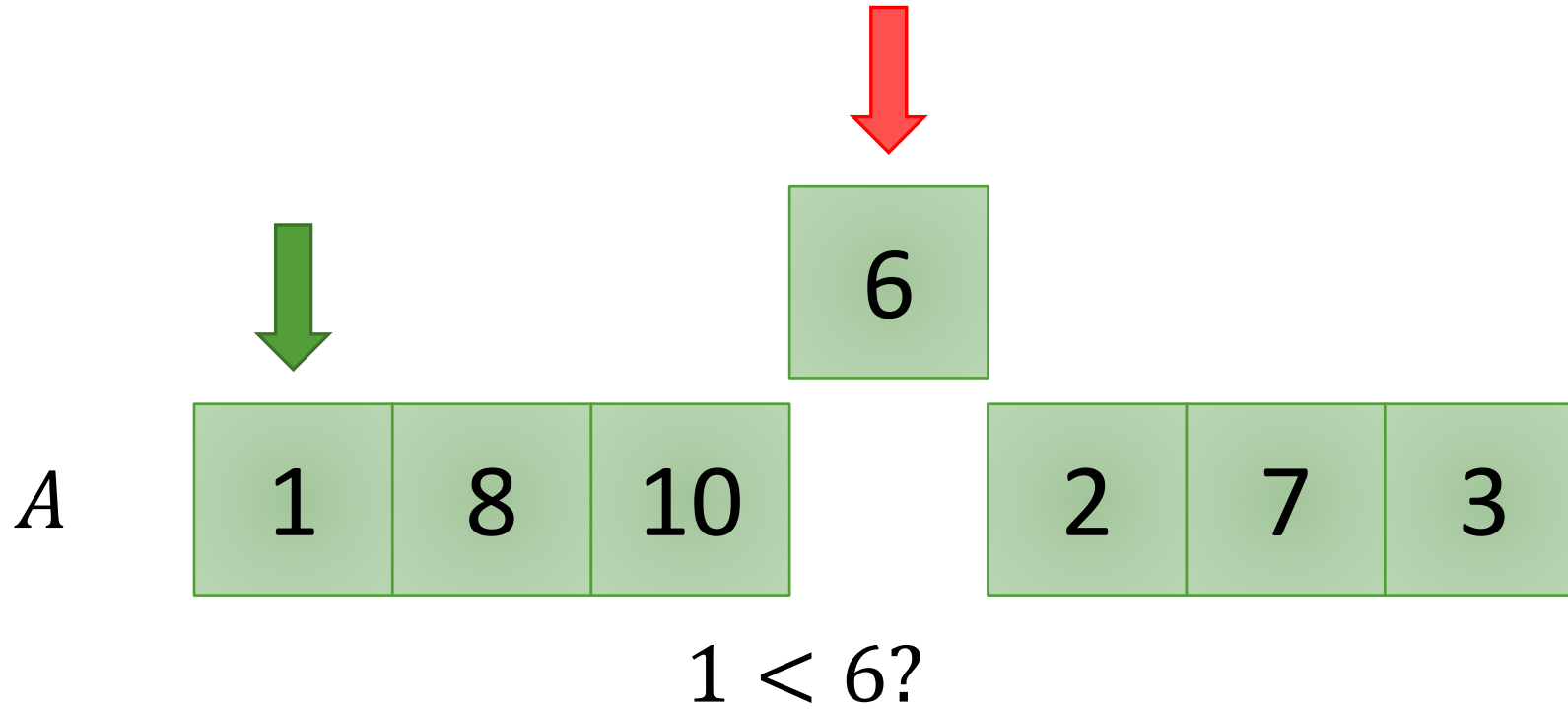
Insertion Sort



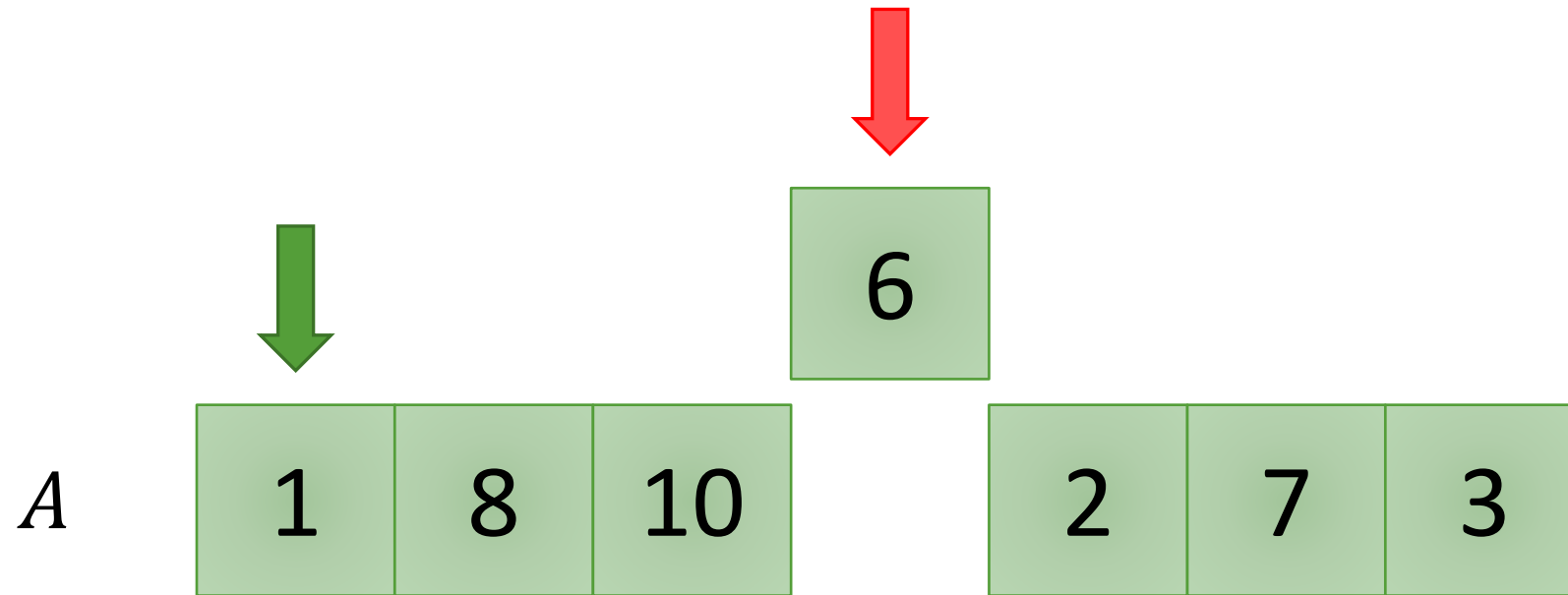
Insertion Sort



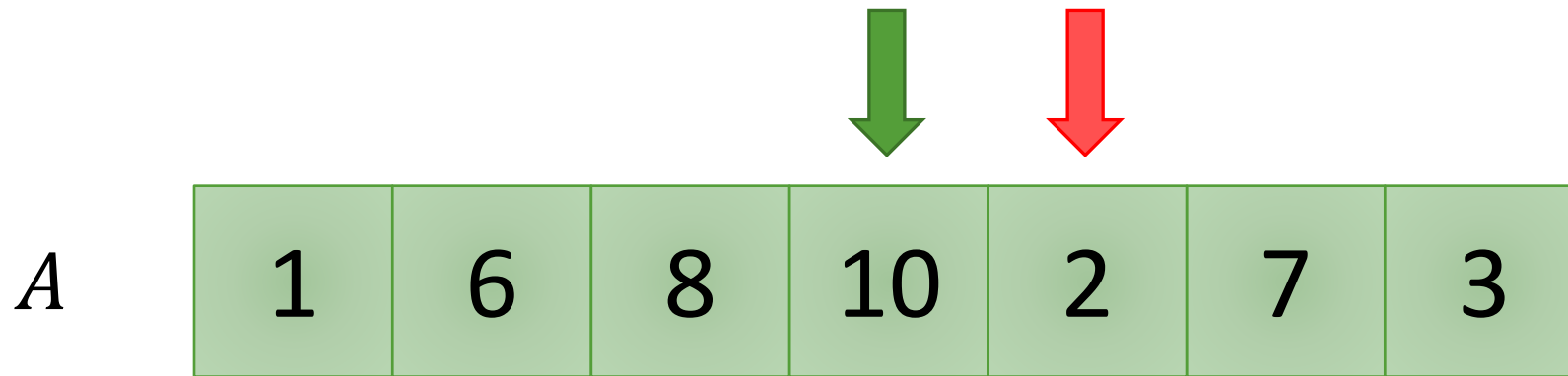
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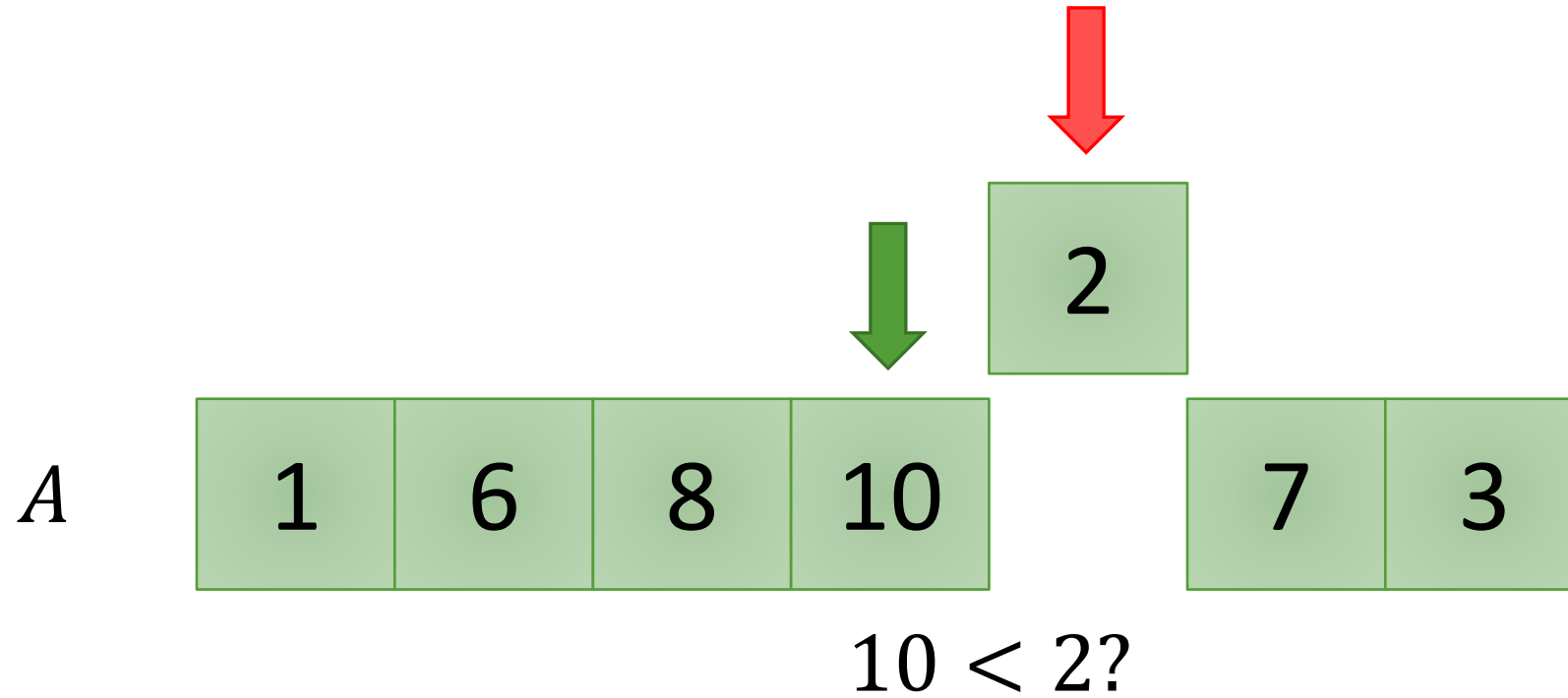
Insertion Sort



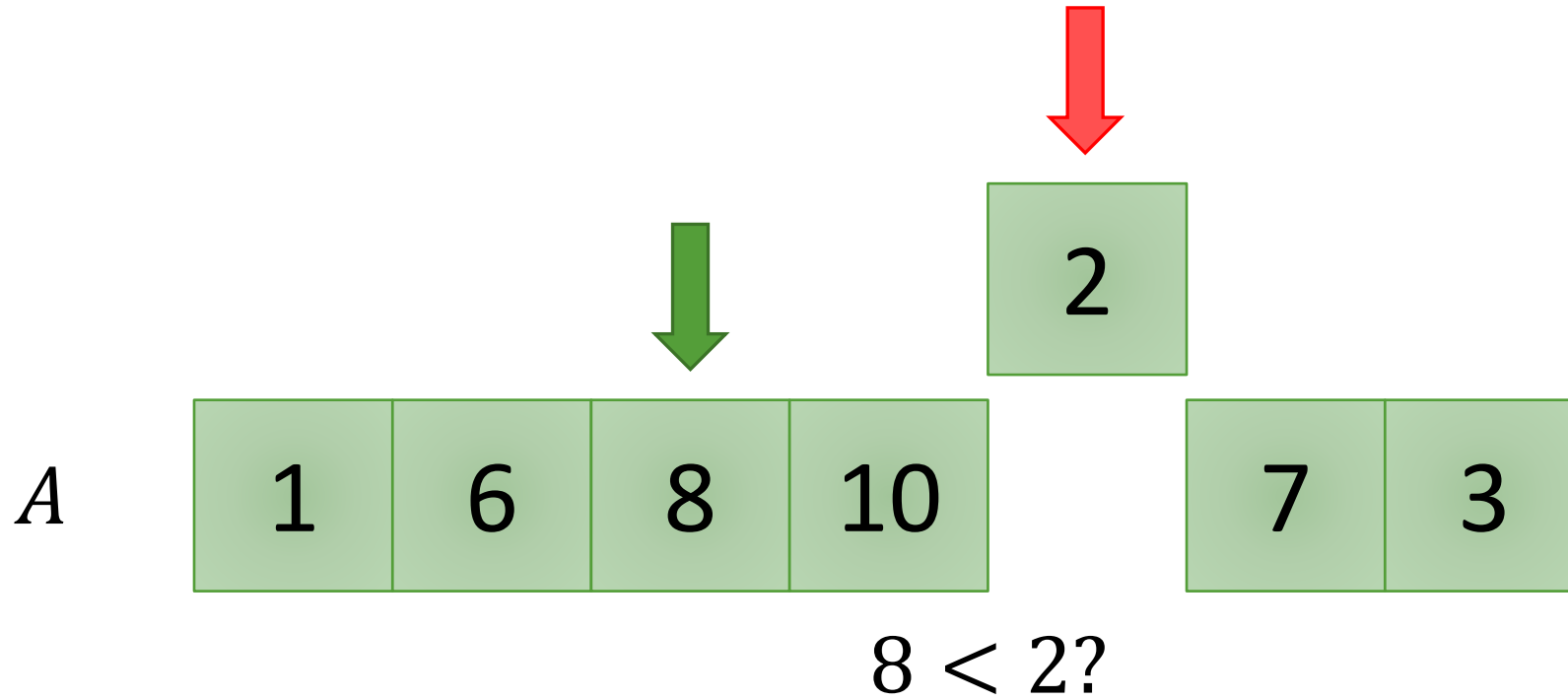
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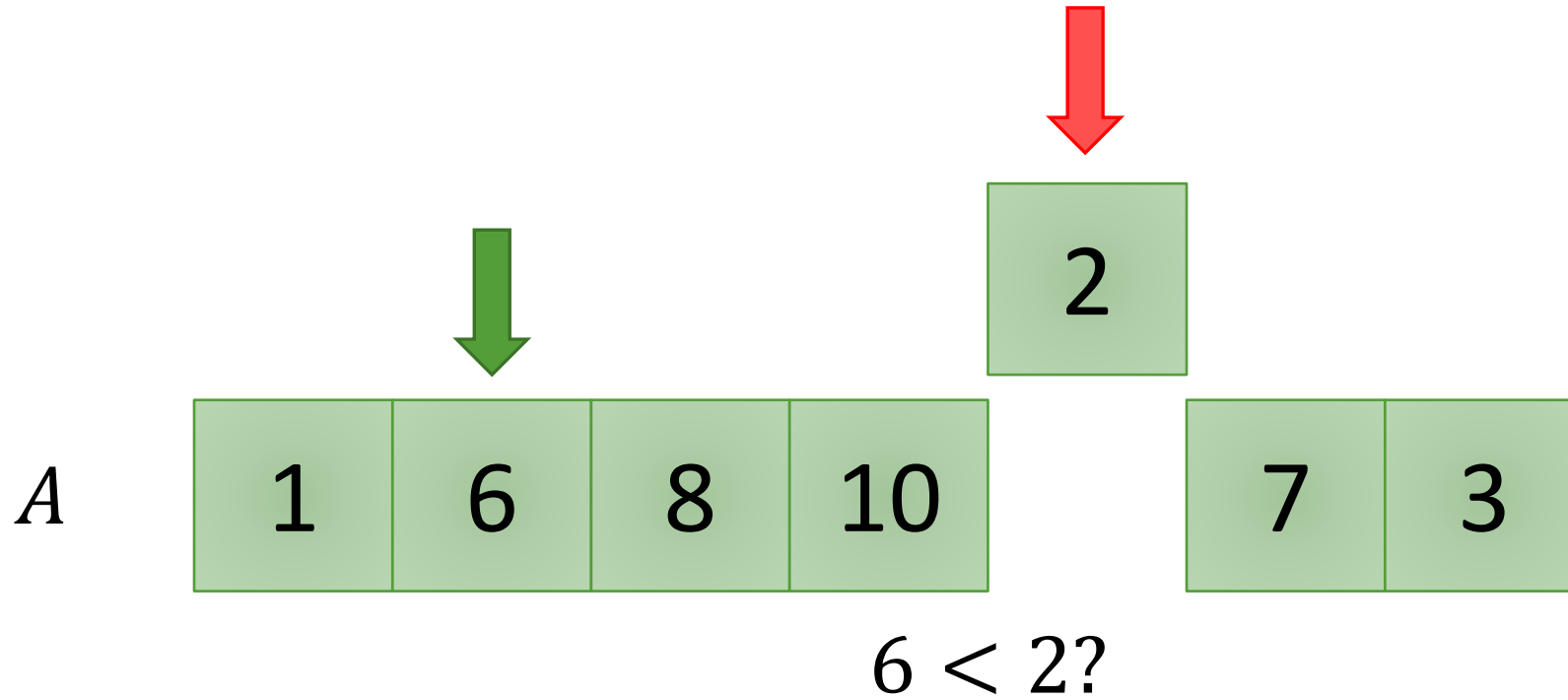
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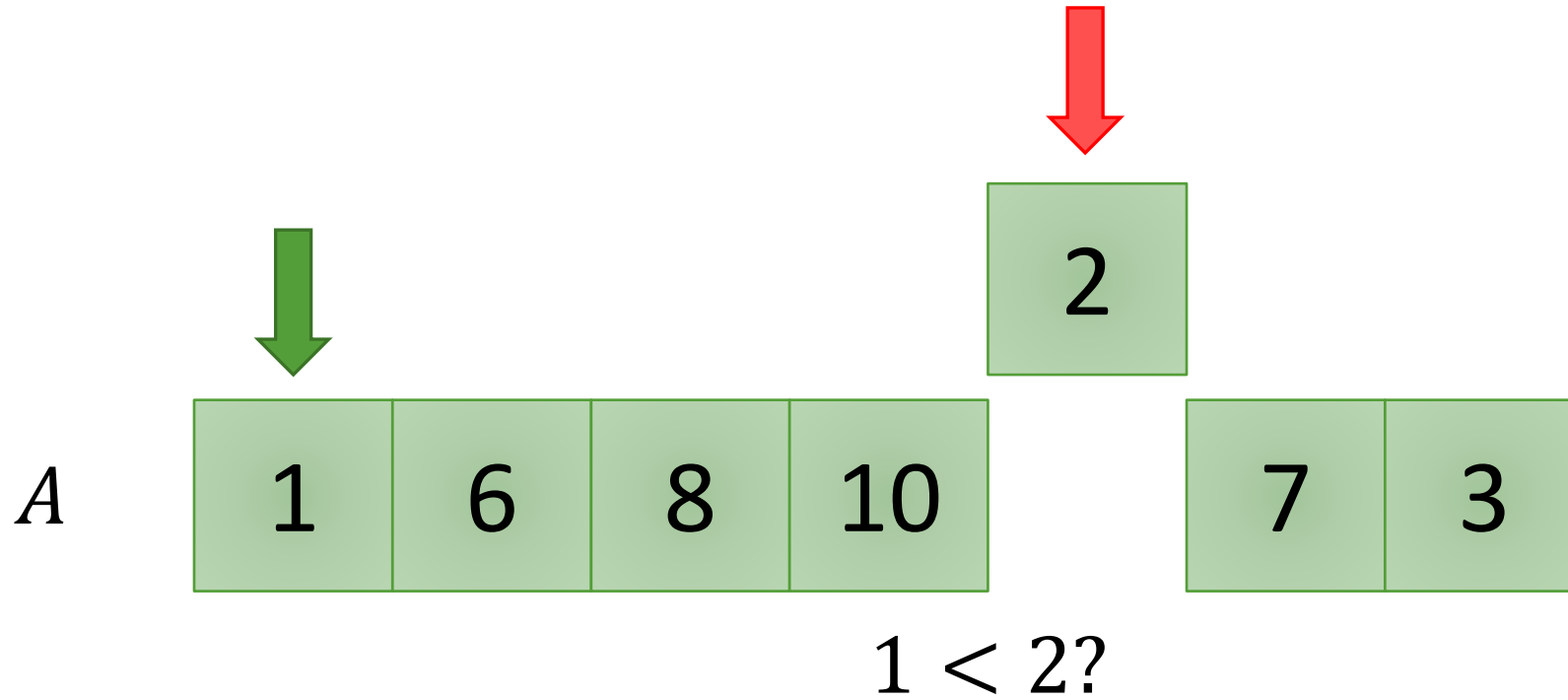
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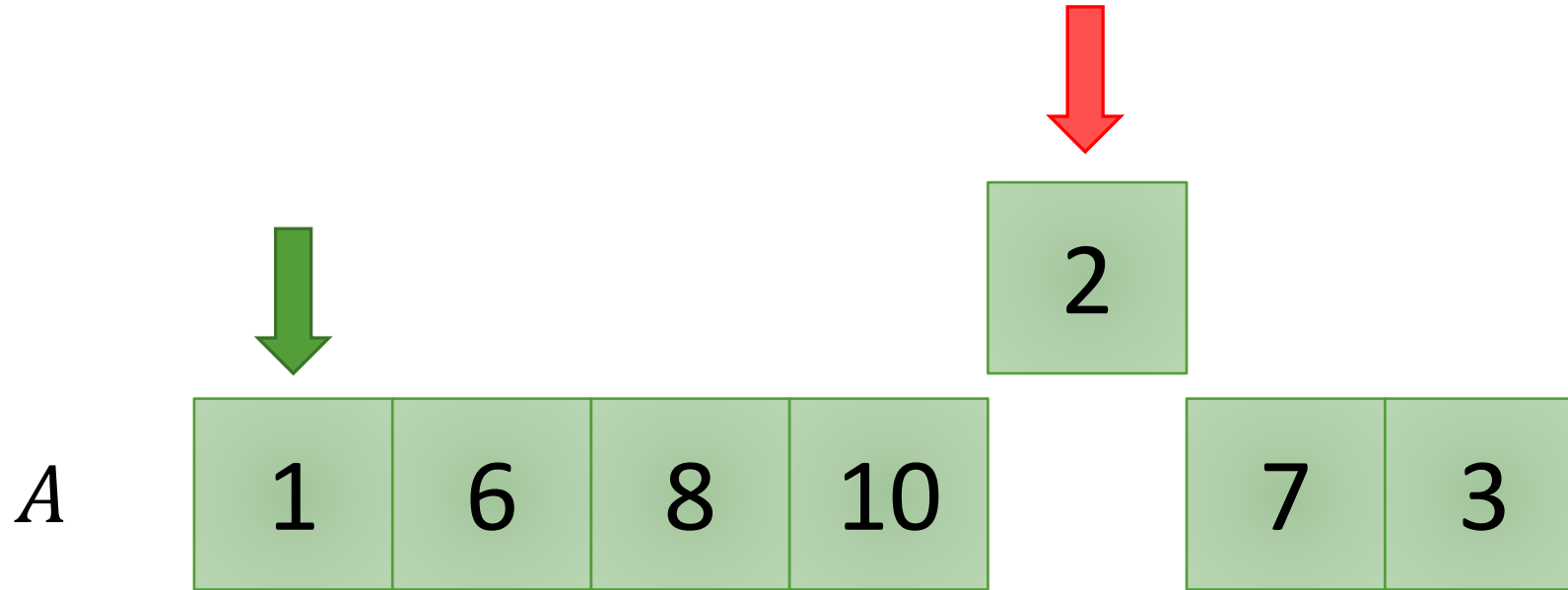
Insertion Sort



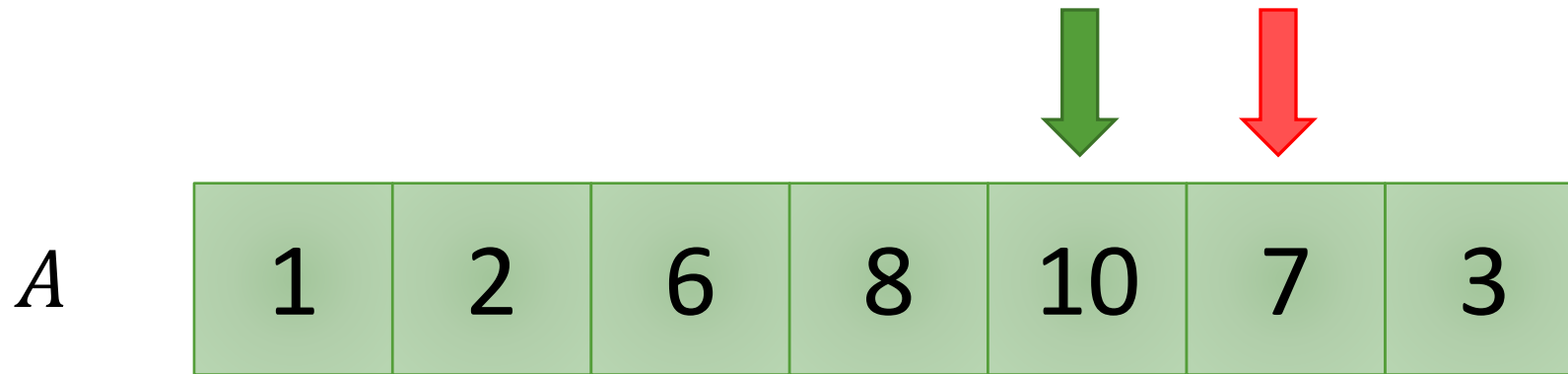
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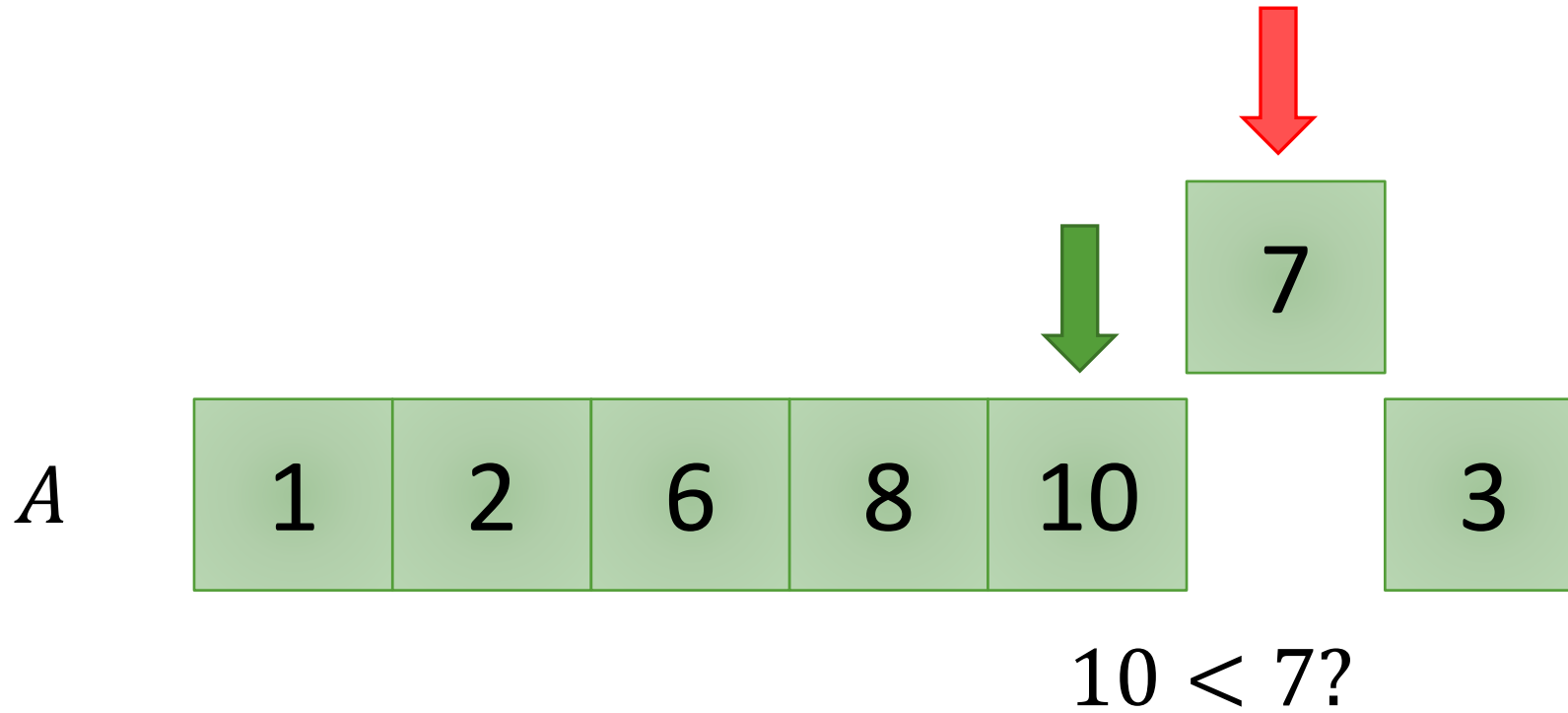
Insertion Sort



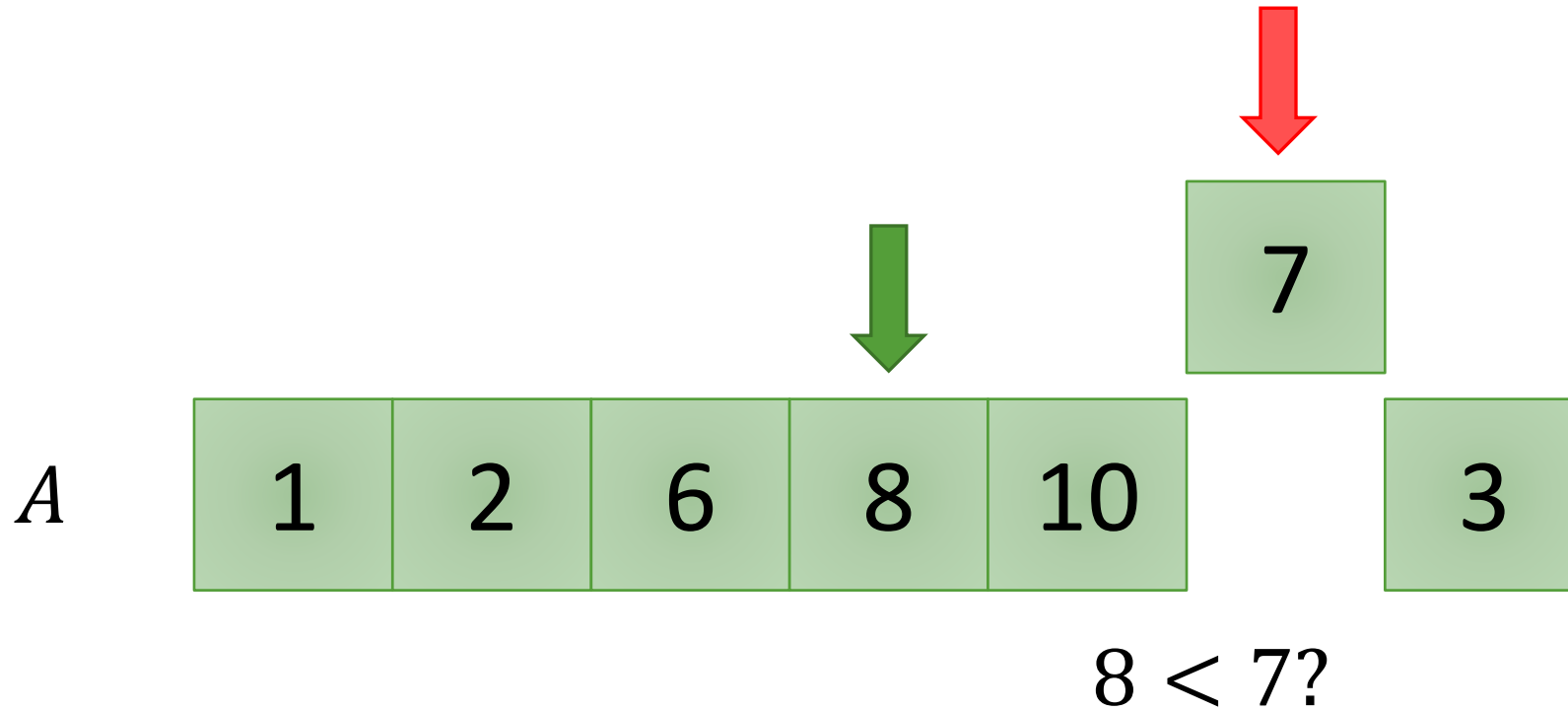
Insertion Sort



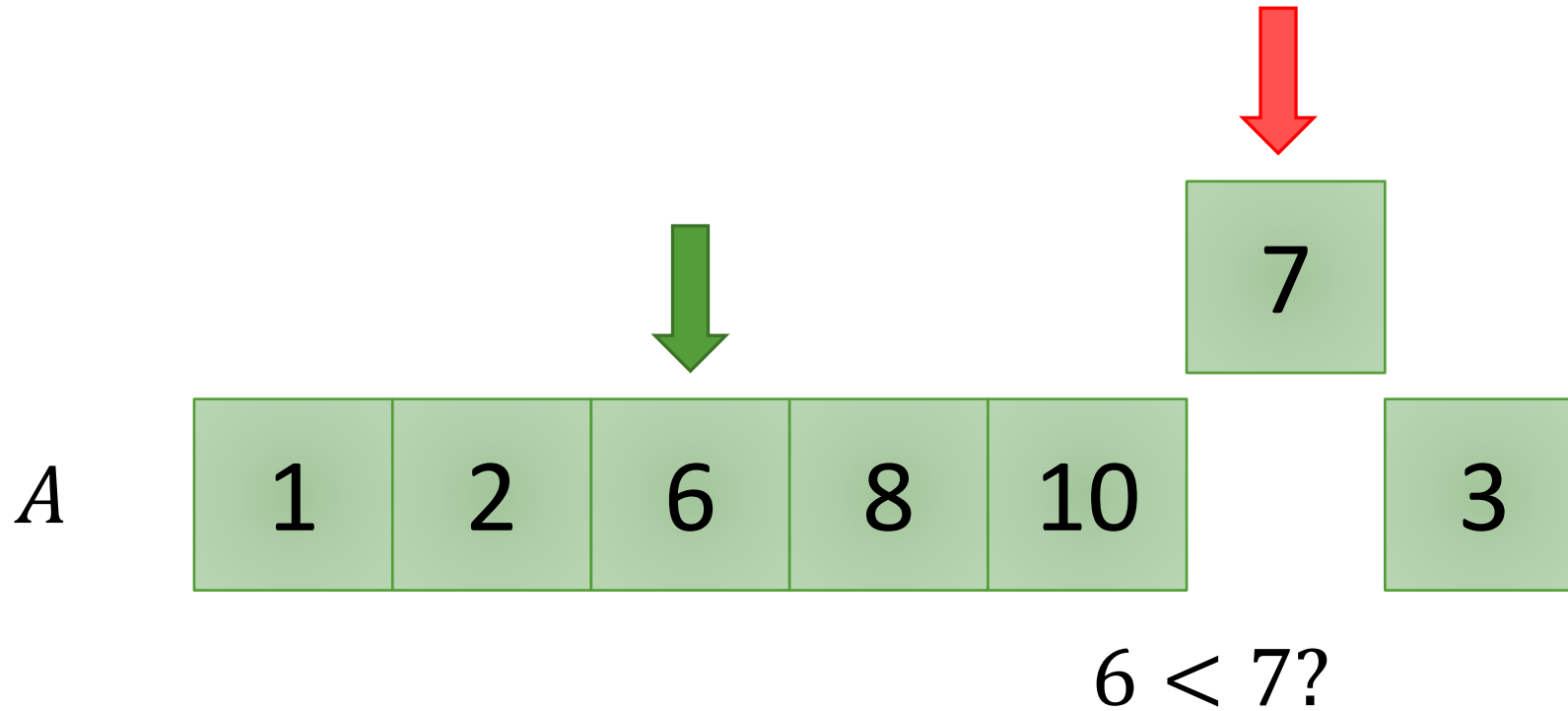
Insertion Sort



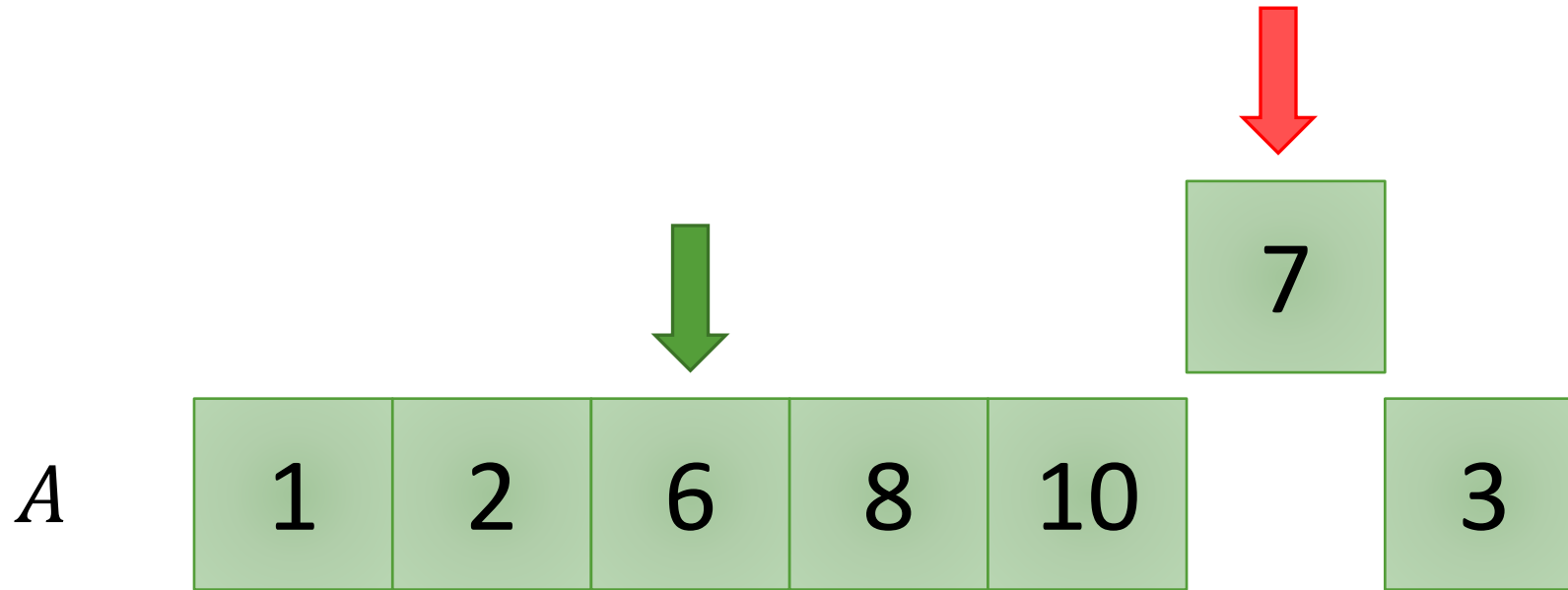
Insertion Sort



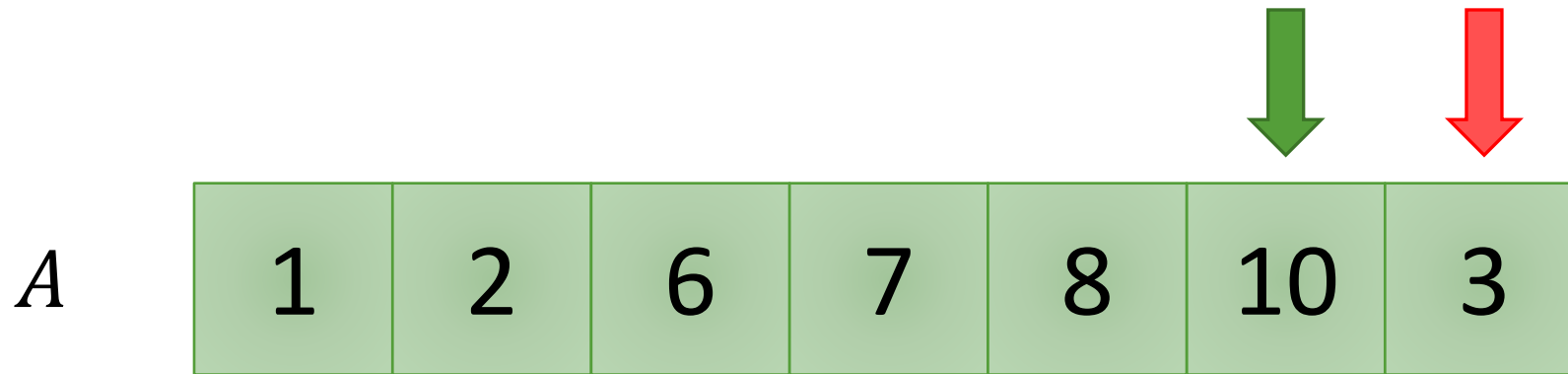
Insertion Sort



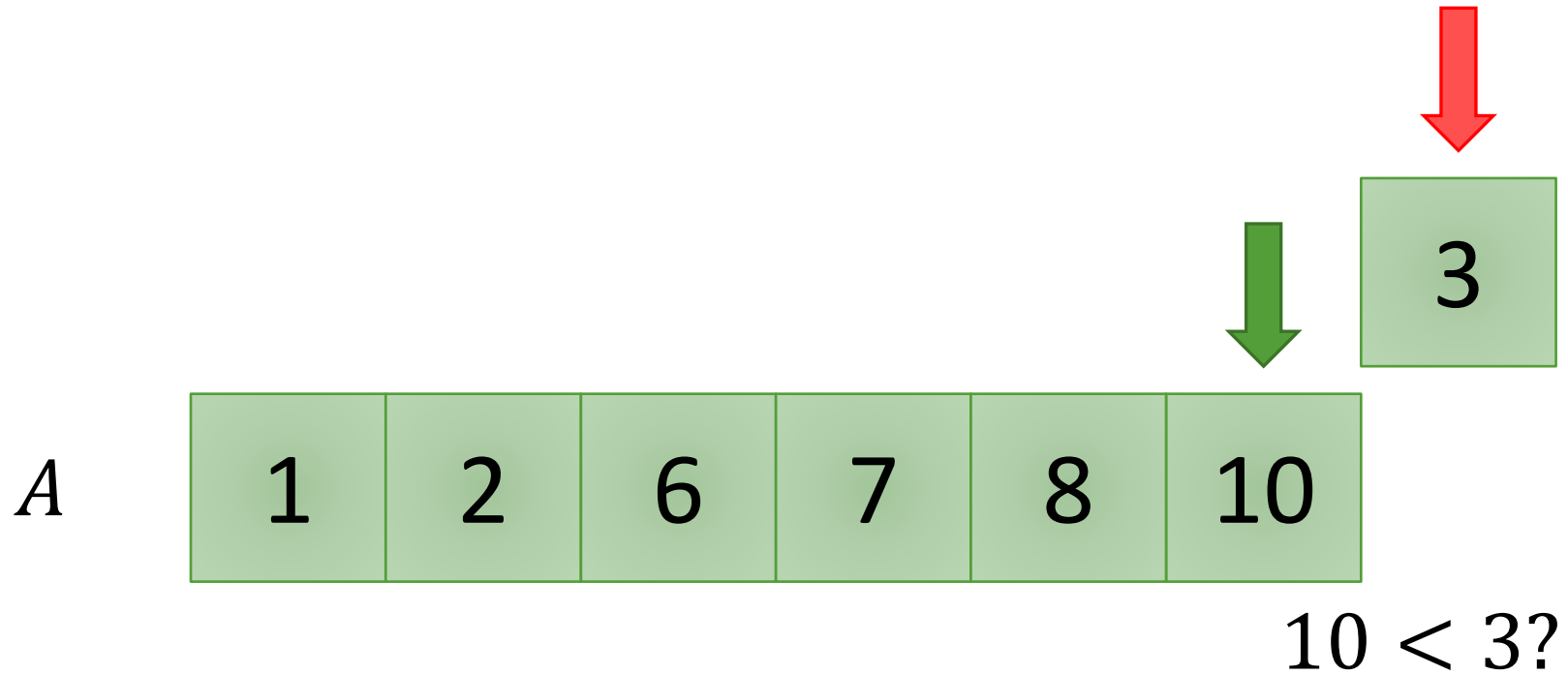
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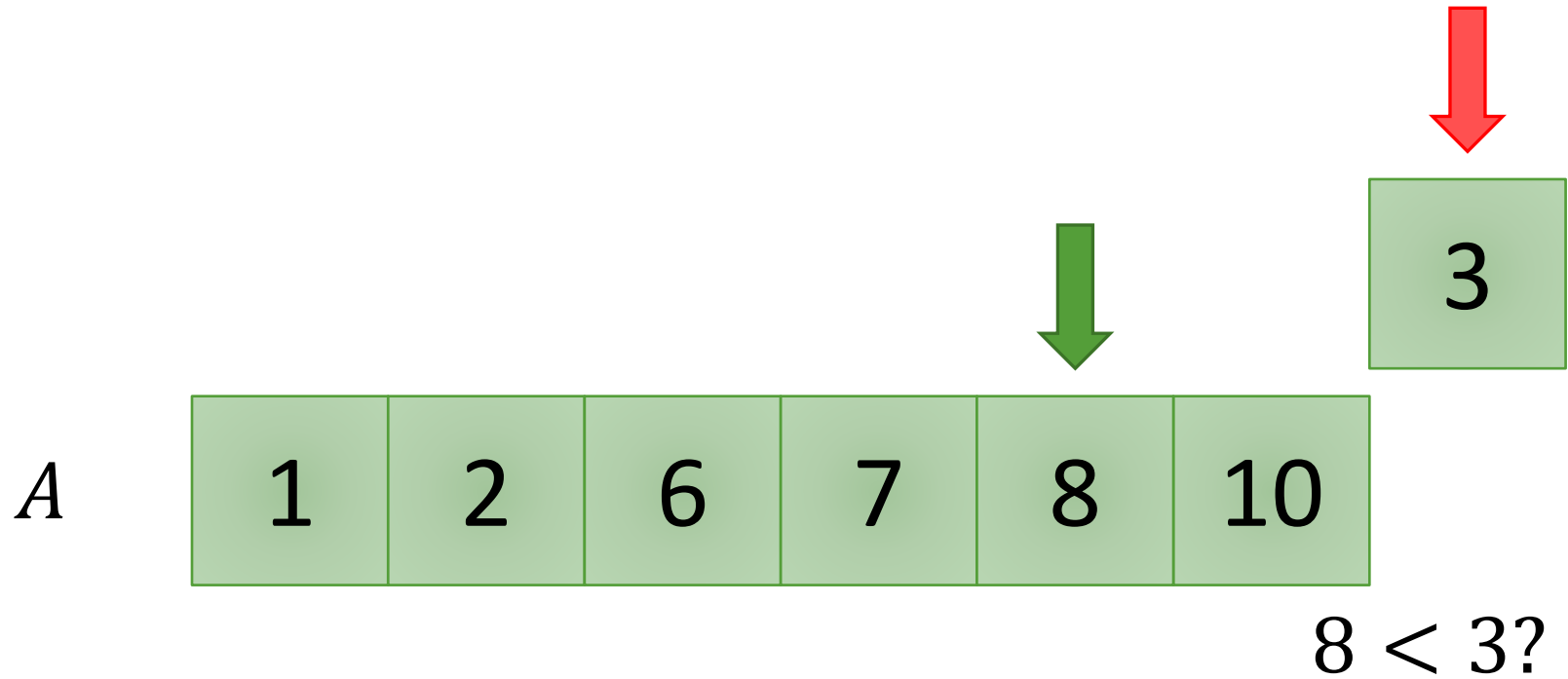
Insertion Sort



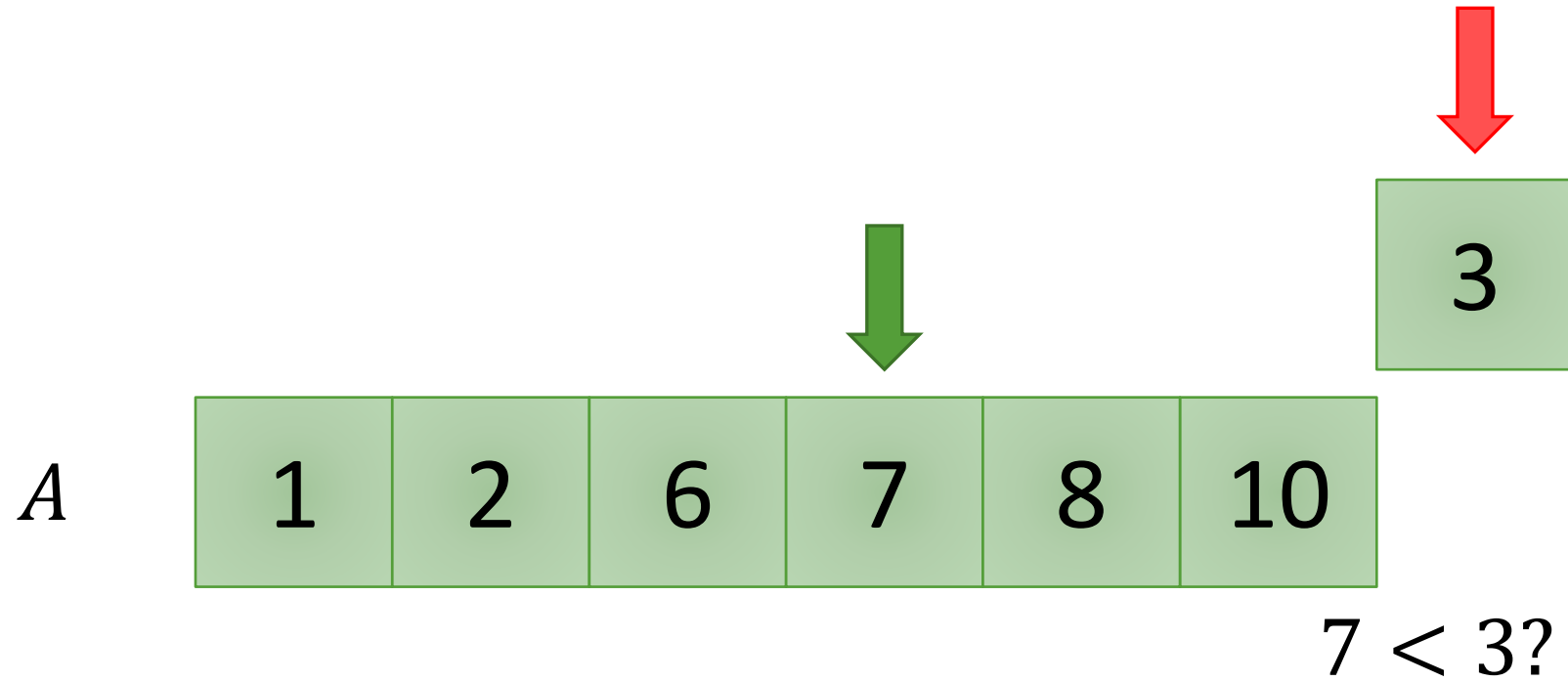
Insertion Sort



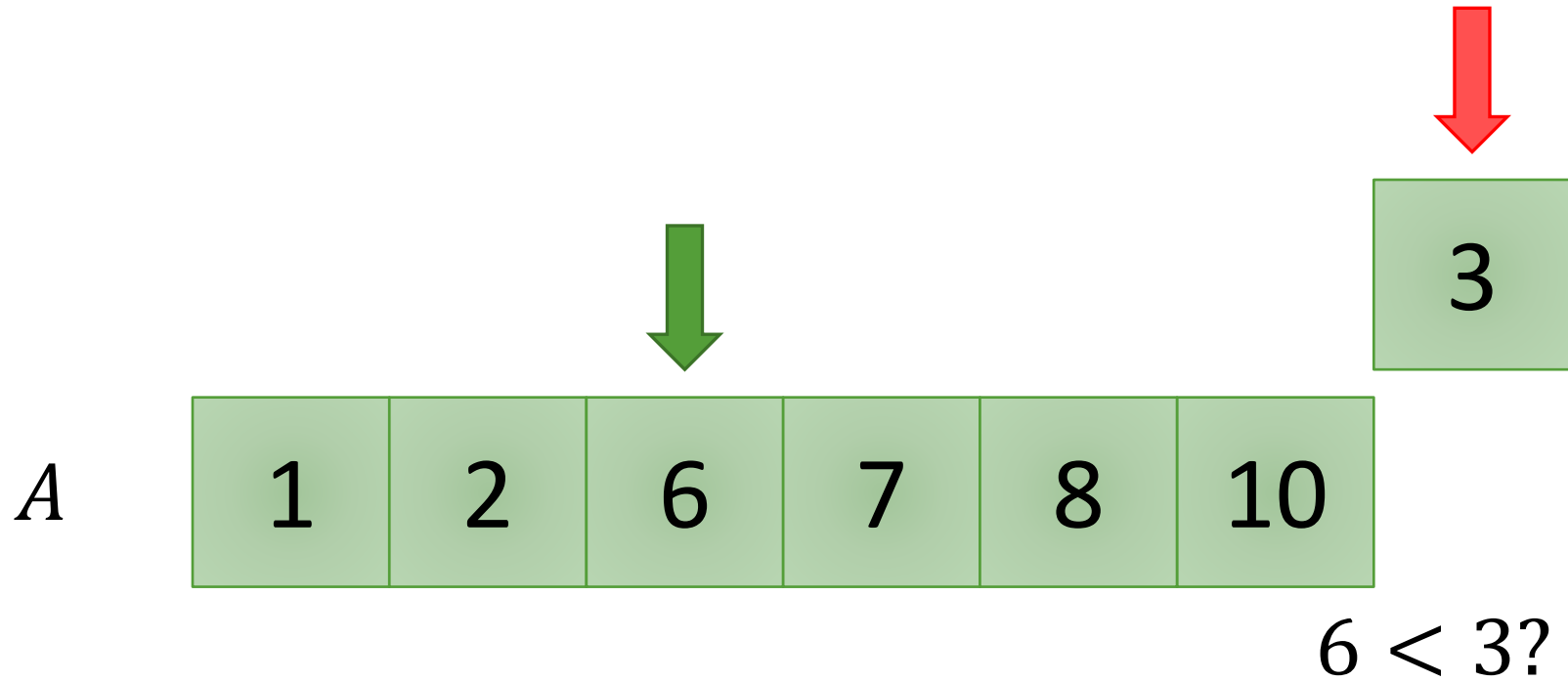
Insertion Sort



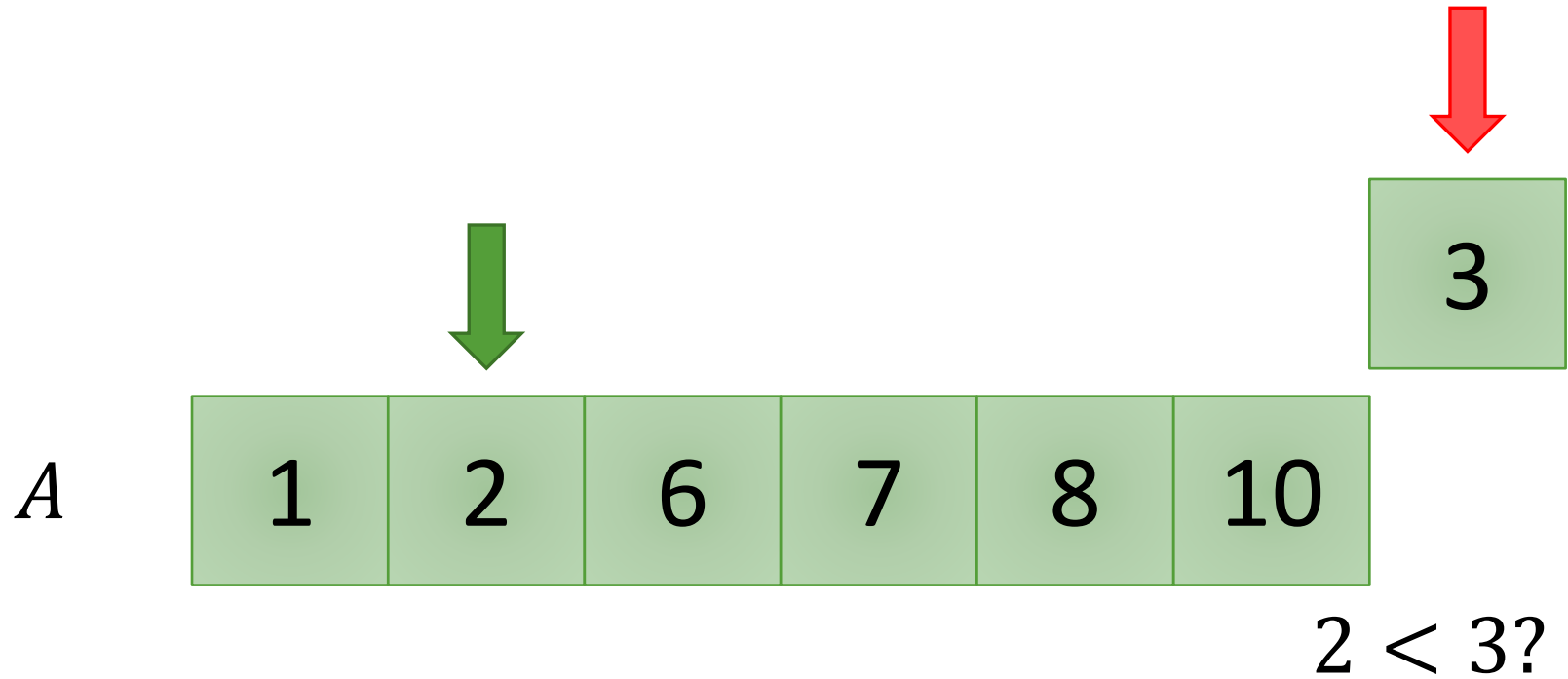
Insertion Sort



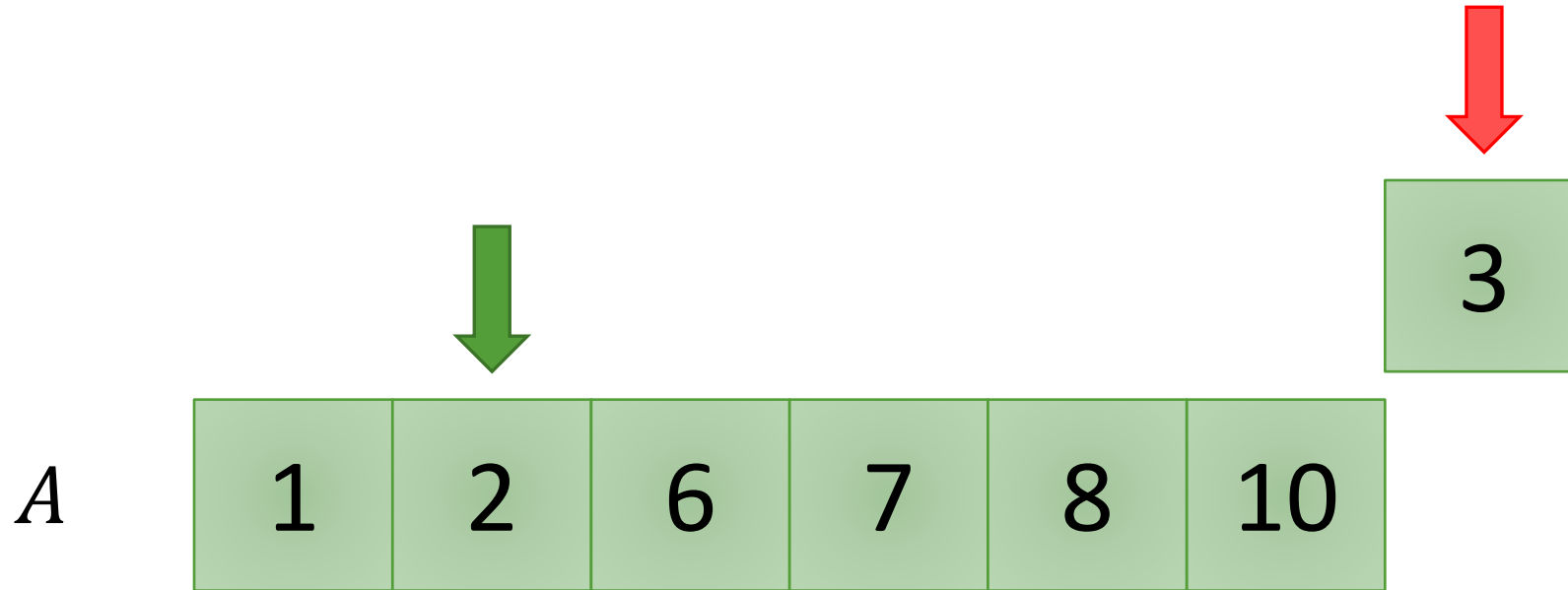
Insertion Sort



Insertion Sort



Insertion Sort



Insertion Sort

A

1	2	3	6	7	8	10
---	---	---	---	---	---	----

Pseudocode

$A[1]$ $A[2]$... \dots $A[n]$

Insertion_Sort(A, n):



Pseudocode

$A[1]$ $A[2]$... $A[n]$

Insertion_Sort(A, n):

for $j = 2$ **to** n

$temp = A[j]$

$i = j - 1$

while $i \geq 1$ and $A[i] > temp$

$A[i + 1] = A[i]$

$i = i - 1$

$A[i + 1] = temp$



Pseudocode

$i = 4$ $j = 5$

Insertion_Sort(A, n):

for $j = 2$ **to** n

temp = $A[j]$

$i = j - 1$

while $i \geq 1$ and $A[i] > temp$

$A[i + 1] = A[i]$

$i = i - 1$

$A[i + 1] = temp$



temp

Pseudocode

$i = 4$ $j = 5$



Insertion_Sort(A, n):

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$temp = A[j]$

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while $i \geq 1$ and $A[i] > temp$

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$i = i - 1$

$A[i + 1] = temp$

1	6	8	10	10	7	3
---	---	---	----	----	---	---

2

$temp$

Pseudocode

Insertion_Sort(A, n):

for $j = 2$ **to** n

$temp = A[j]$

$i = j - 1$

while $i \geq 1$ and $A[i] > temp$

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$temp$

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$A[i + 1] = temp$

$i = 1$



$j = 5$



1	6	6	8	10	7	3
---	---	---	---	----	---	---

2

$temp$

Pseudocode

Insertion_Sort(A, n):

for $j = 2$ **to** n

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$i = j - 1$

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$i = i - 1$

$A[i + 1] = temp$

$i = 1$



$j = 5$



1	2	6	8	10	7	3
---	---	---	---	----	---	---

2

$temp$

Analyzing Insertion Sort

- We need to show:
 1. That the algorithm is **correct** for ANY input. That is, it will always output a sorted array.
 2. The **running time** of the algorithm.
- This is the case for all algorithms that we will study.
 - Except we will often put more emphasis on finding the running time

“Testing can only show the **presence** of errors, not their **absence**”
- E. W. Dijkstra

Analyzing Correctness

- How do you prove correctness?
- What were you taught to do in your Software Engineering class?
 - Inspections
 - Walkthroughs
 - Testing
 - Unit Testing
 - Integration Testing
 - Functional/System Testing
- But testing is inconclusive and not a proof that there are no errors.

Analyzing Correctness

- **Definition:** An algorithm is said to be **correct** if it completely satisfies the specification it is meant to accomplish with respect to the problem it is solving and terminates with the expected output.
- We will use **formal verification** to prove that algorithms satisfy certain **properties** or **invariants** that ensure correctness.
- Such a proof is called a **Loop Invariant Proof**.

Correctness: Loop Invariant Proof

- To **prove** that an algorithm is **correct** for ANY input, we will need to show that it is correct (so far) for EVERY iteration of the loop:
 1. Formulate a **loop invariant** that defines what correctness means
 2. **Initialization**: Show that the loop invariant is true at the start of the **first** iteration
 3. **Maintenance**: Assume loop invariant is true at the start of iteration j . Then prove it is true at the start of **iteration $j + 1$**
 4. **Termination**: Show that, when the loop terminates, the algorithm outputs the correct answer.

Analyzing Correctness

Insertion_Sort(A, n):

for $j = 2$ **to** n

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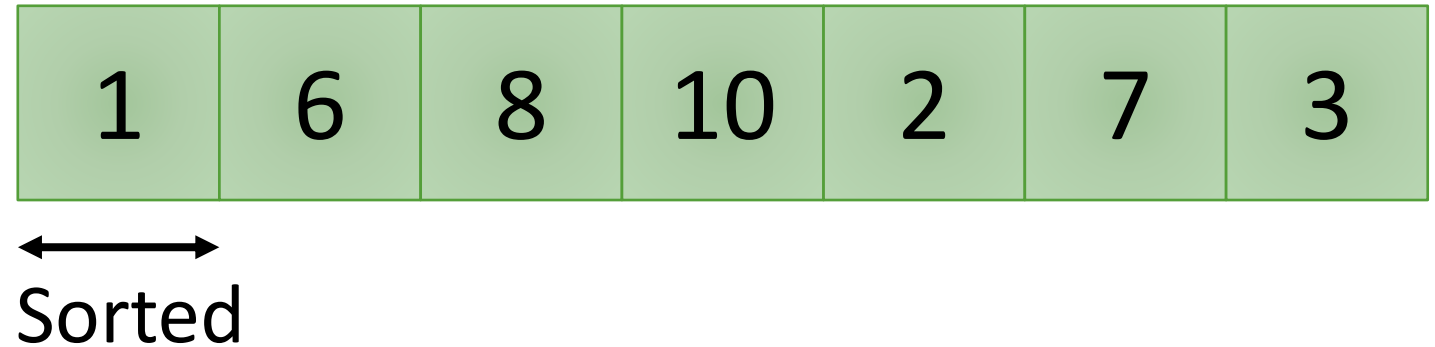
Loop Invariant:

At the start of iteration j of the **for** loop, subarray $A[1, \dots, j - 1]$ is sorted.

Analyzing Correctness



- ✓ **Initialization:** Show that the loop invariant is true in the **first** iteration
- Is the loop invariant true in the first iteration?
- Yes. Proof: At the start of the first iteration ($j = 2$), the array $A[1, \dots, 1] = A[1]$ is sorted (trivially)

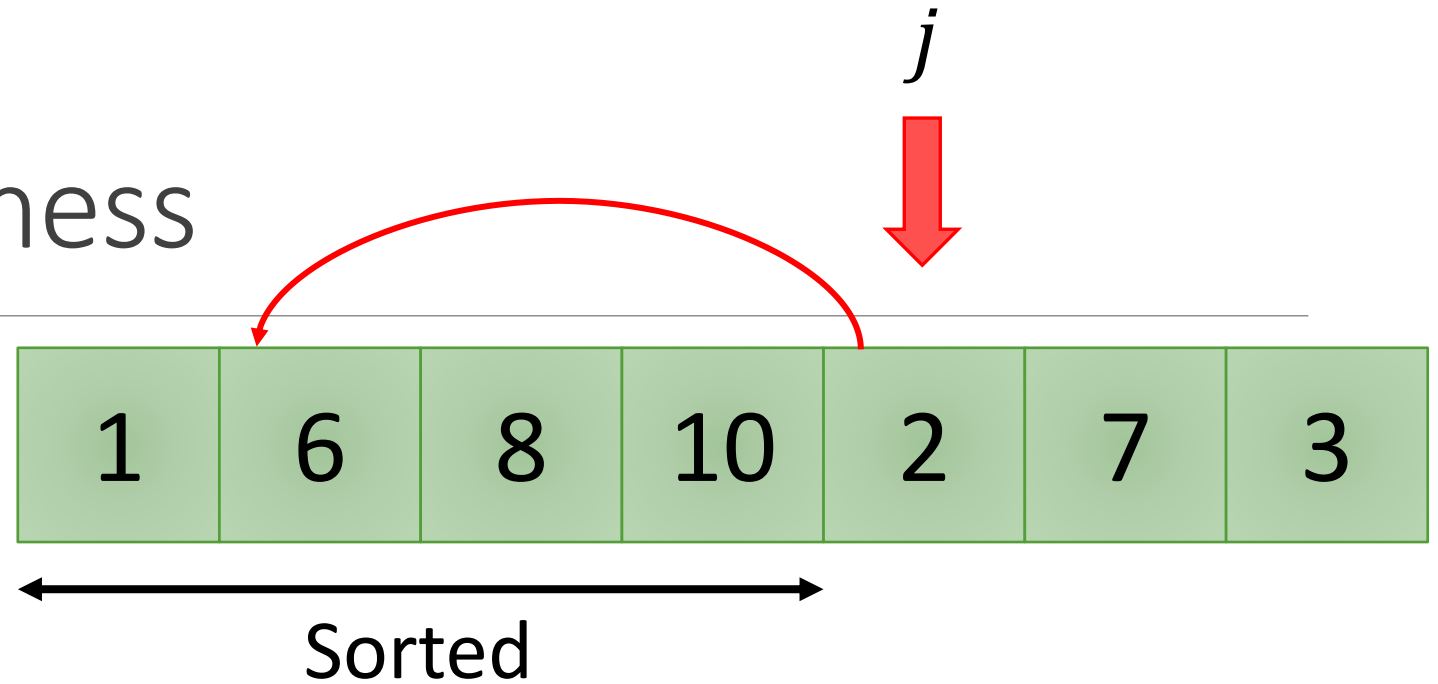


Loop Invariant:

At the start of iteration j of the **for** loop, subarray $A[1, \dots, j - 1]$ is sorted.

Analyzing Correctness

- ✓ **Maintenance:** Assume loop invariant is true at the start of iteration j . Then prove it is true at the start of **iteration $j + 1$**



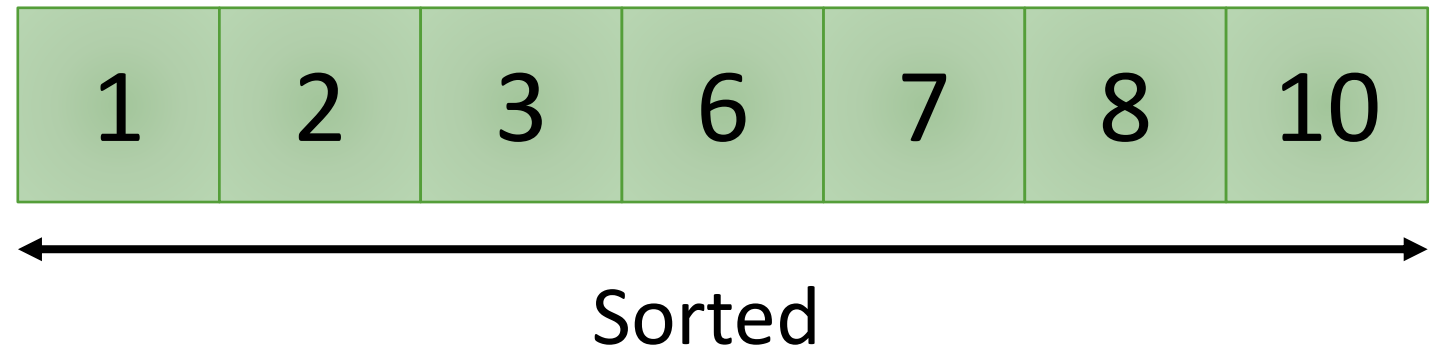
- **Proof:** Informally, if we assume that loop invariant is true at start of iteration j then $A[1, \dots, j - 1]$ is sorted.
 - Then, at start of iteration $j + 1$ the element $A[j]$ will be inserted in the proper location within $A[1, \dots, j]$ by the inner while loop so $A[1, \dots, j]$ will also be sorted

Loop Invariant:

At the start of iteration j of the **for** loop, subarray $A[1, \dots, j - 1]$ is sorted.

Analyzing Correctness

- ✓ **Termination:** Show that, when the loop terminates, the algorithm outputs the correct answer.



- Proof: The loop terminates when $j = n + 1$. At this point, the loop invariant states that the subarray $A[1, \dots, (n + 1) - 1]$ is sorted, which is the entire array – so the output is correct.

Loop Invariant:

At the start of iteration j of the **for** loop, subarray $A[1, \dots, j - 1]$ is sorted.

Analyzing running time

- The running time T must be a **function of the input**; there are inputs for which the algorithm is slow and others for which the algorithm is fast.
- We parametrize the running time of an algorithm with the **input size**. i.e., $T(n)$ where n is the **input size**.

Types of Analyses

- Worst-case:
 - $T(n)$ = **MAXIMUM** running time of the algorithm on input of size n
- Average-case:
 - $T(n)$ = **EXPECTED** running time of the algorithm over **all** inputs of size n
- Best-case:
 - $T(n)$ = **MINIMUM** running time of the algorithm over input of size n

Types of Analyses

- Worst-case:
 - $T(n)$ = **MAXIMUM** running time of the algorithm on input of size n
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 - $T(n)$ = **EXPECTED** running time of the algorithm over **all** inputs of size n
- Best-case:
 - $T(n)$ = **MINIMUM** running time of the algorithm over input of size n

Worst-case running time

- What is insertion's sort worst-case running time?
 - It depends!
- Depends on:
 - CPU clock speed
 - Memory speed
 - Cache size
 - Etc.
- Need a **machine-independent** way of measuring time!

Asymptotic Analysis

- Main Idea:
 - Identify the **primitive** operations and the **data structure** used
 - Ignore machine-dependent **constants** and factors
 - Examine the **growth** of $T(n)$ as $n \rightarrow \infty$

Θ – Notation (informally)

1. Drop low-order terms
2. Ignore leading constants

- E.g. $T(n) = 22n^2 + 41n + 12 = \Theta(n^2)$
- E.g. $T(n) = 4n^2 + 3n^5 + 2n^8 = \Theta(n^8)$

Worst-case running time of insertion sort

Insertion_Sort(A, n):

for $j = 1$ **to** n

$temp = A[j]$

$i = j - 1$

while $i > 0$ and $A[i] > temp$

$A[i + 1] = A[i]$

$i = i - 1$

$A[i + 1] = temp$

10

8

7

6

3

2

1

1. Data structure? **Array**
2. Primitive operations?

Worst-case running time of insertion sort

Insertion_Sort(A, n):

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10

8

7

6

3

2

1

1. Data structure? Array
2. Primitive operations?
Compare, **Shift**, **Assign**

Worst-case running time of insertion sort

Insertion_Sort(A, n):

for $j = 1$ **to** n

$temp = A[j]$

$i = j - 1$

while $i > 0$ and $A[i] > temp$

$A[i + 1] = A[i]$

$i = i - 1$

$A[i + 1] = temp$

10	8	7	6	3	2	1
----	---	---	---	---	---	---

j	#Compares	#Shifts	#Assign	$T_j(n)$
1	0	0	1	0
2	1	1	1	2+1
3	2	2	1	4+1
...
n	$n-1$	$n-1$	1	$2(n-1)+1$

$$T(n) = \sum_j T_j(n) = \sum_j 2(j-1) + 1 = \Theta(n^2)$$

Best-case running time of insertion sort

Insertion_Sort(A, n):

for $j = 1$ **to** n

$temp = A[j]$

$i = j - 1$

while $i > 0$ and $A[i] > temp$

$A[i + 1] = A[i]$

$i = i - 1$

$A[i + 1] = temp$

1	2	3	6	7	8	10
---	---	---	---	---	---	----

j	#Compares	#Shifts	#Assign	$T_j(n)$
1	0	0	1	0
2	1	0	1	1+1
3	1	0	1	1+1
...
n	1	0	1	1+1

$$T(n) = \sum_j T_j(n) = \sum_j 2 = \Theta(n)$$

Asymptotic Analysis

- What can we say about the relative performance of any two algorithms using Θ –notation?
- A $\Theta(n^2)$ algorithm is always faster than a $\Theta(n^3)$ algorithm **for large enough n** .
 - We say the $\Theta(n^2)$ algorithm is asymptotically faster than the $\Theta(n^3)$ one.
- Though sometimes, in practice, we might favor asymptotically slower algorithms if we only care about “smaller” values of n

Try this example: Finding the Maximum

```
Find_Max( $A, n$ ):  
 $m = A[1]$   
for  $j = 2$  to  $n$   
    if  $A[j] > m$  then  
         $m = A[j]$   
return  $m$ 
```

- Prove **correctness**
 - Define loop invariant
 - Initialization
 - Maintenance
 - Termination
- Find the **running time**

Finding the Maximum: Correctness

- **Loop invariant:** At the start of iteration j , m is the maximum of $A[1, \dots, j - 1]$
- **Initialization:** At start of first iteration $j = 2$, it is true that $m = A[1]$ is clearly the maximum of subarray $A[1, \dots, 1] = A[1]$

```
Find_Max( $A, n$ ):  
 $m = A[1]$   
for  $j = 2$  to  $n$   
    if  $A[j] > m$  then  
         $m = A[j]$   
return  $m$ 
```

Finding the Maximum: Correctness

- **Loop invariant:** At the start of iteration j , m is the maximum of $A[1, \dots, j - 1]$
- **Maintenance:** Assume m is the maximum of $A[1, \dots, j - 1]$ at start of iteration j
- Then at iteration $j + 1$, either m stays the same or $A[j + 1]$ is the new maximum. So m now becomes the maximum of $A[1, \dots, j]$

```
Find_Max( $A, n$ ):  
 $m = A[1]$   
for  $j = 2$  to  $n$   
    if  $A[j] > m$  then  
         $m = A[j]$   
return  $m$ 
```

Finding the Maximum: Correctness

- **Loop invariant:** At the start of iteration j , m is the maximum of $A[1, \dots, j - 1]$
- **Termination:** Loop terminates when $j = n + 1$ at which point the loop invariant states that m is the maximum of $A[1, \dots, (n + 1) - 1] = A[1, \dots, n]$ which is indeed what is required of this algorithm

```
Find_Max( $A, n$ ):  
 $m = A[1]$   
for  $j = 2$  to  $n$   
    if  $A[j] > m$  then  
         $m = A[j]$   
return  $m$ 
```