

Quick Sort: Analysis + Linear Time Sorting

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CSC 212

Announcements

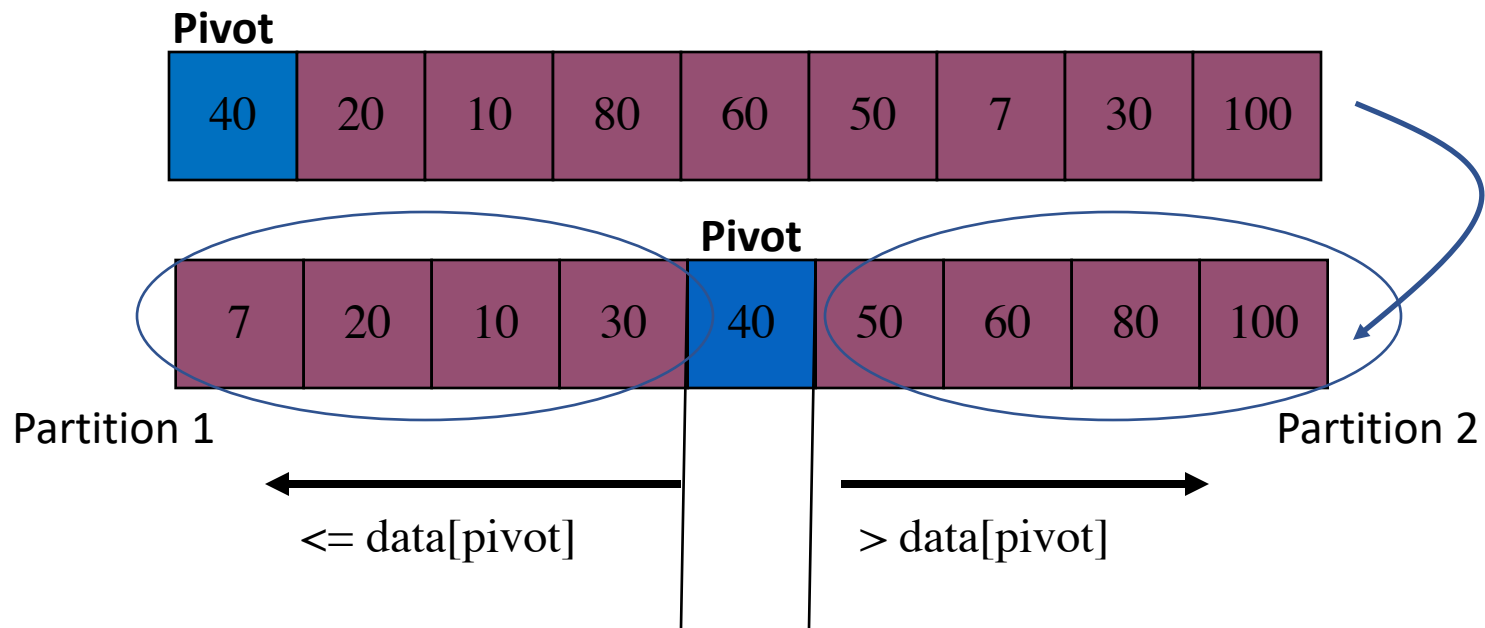
- No class next Tuesday Oct 15
- Quiz 3 will be Tuesday, Oct 22
 - It will cover material from the lectures of Oct 8, Oct 10, and Oct 17.
- There will be no class on Tuesday, Oct 29 and Thursday, Oct 31, as I will be traveling
 - However, **there will be lab that week** on Wednesday and Friday.

Quicksort Algorithm (Recap)

```
QuickSort (A, l, r)
    if r-l+1 == 1
        return
    else
        p = Partition (A, l, r)
        QuickSort (A, l, p)
        QuickSort (A, p+1, r)
```

How to Partition (Recap)

- Given an array *A*
 - Pick one element to use as *pivot*.
 - Partition elements into two sub-arrays:
 - Elements less than or equal to pivot (at index *p*)
 - Elements greater than pivot (at index *p*)

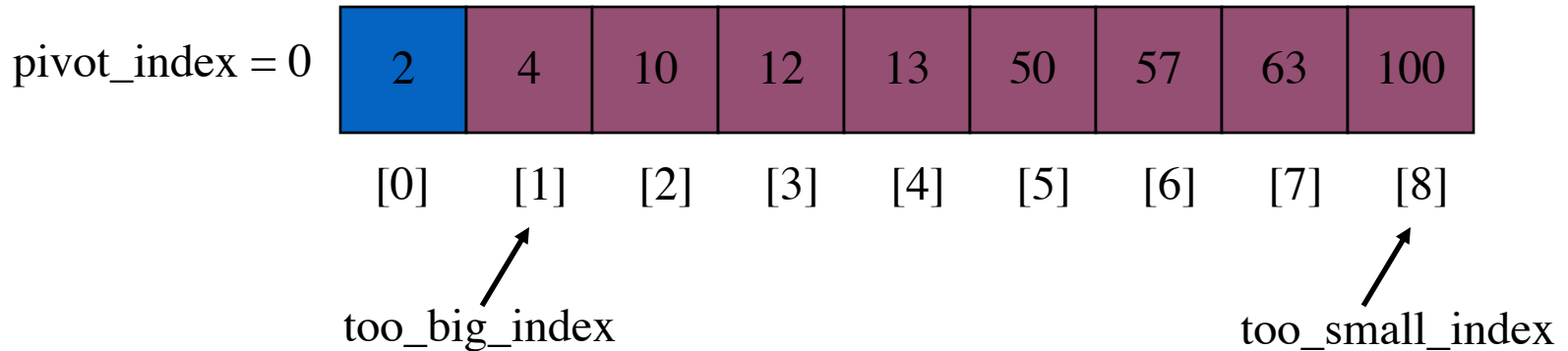


Quicksort Analysis (Recap)

- Assume that keys are random, uniformly distributed.
- What is best case running time?
 - Recursion:
 1. Partition splits array in two sub-arrays of size $n/2$
 2. Quicksort each sub-array
 - Depth of recursion tree? $O(\log_2 n)$
 - Number of accesses in partition? $O(n)$
- Assume that keys are random, uniformly distributed.
- **Best case running time: $O(n \log_2 n)$**

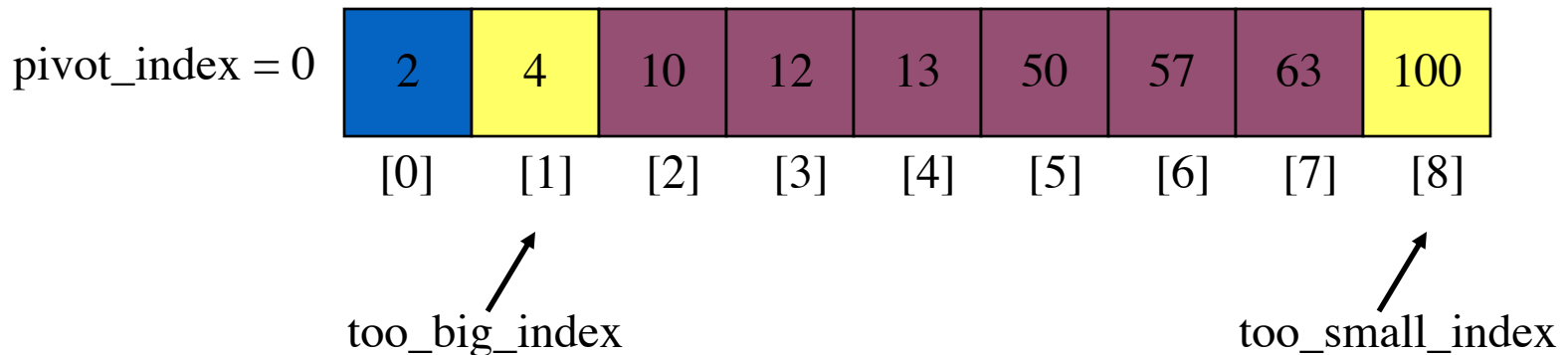
Quicksort: Worst Case

- Assume first element is chosen as pivot.
- Assume we **get array that is already sorted**:



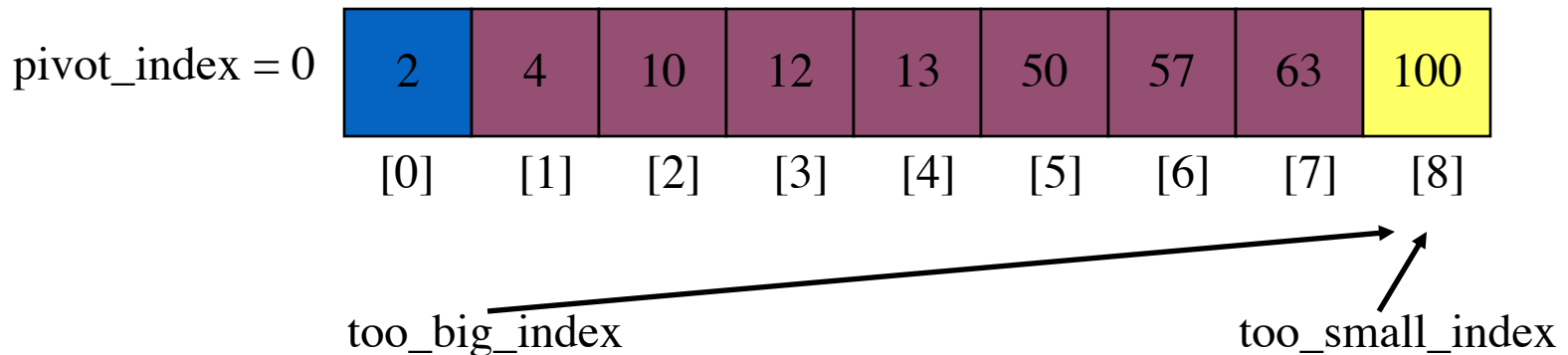
Worst-Case Operation: Example

- 1. **while** data[too_big_index] <= data[pivot]
 too_big_index = too_big_index+1
- 2. **while** data[too_small_index] > data[pivot]
 too_small_index = too_small_index-1
- 3. **if** too_big_index < too_small_index
 swap data[too_big_index] and data[too_small_index]
- 4. **while** too_small_index > too_big_index, go to 1.
- 5. swap data[too_small_index] and data[pivot_index]



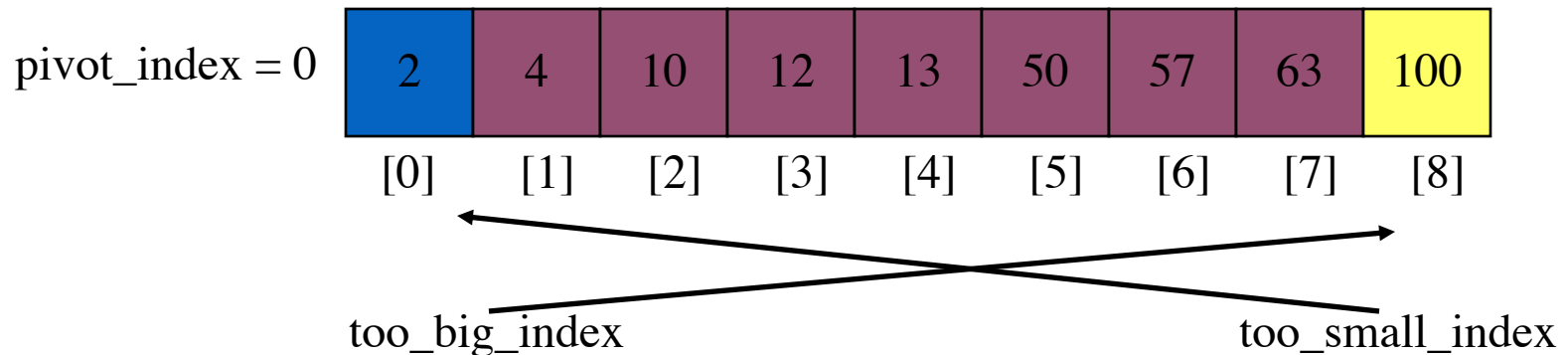
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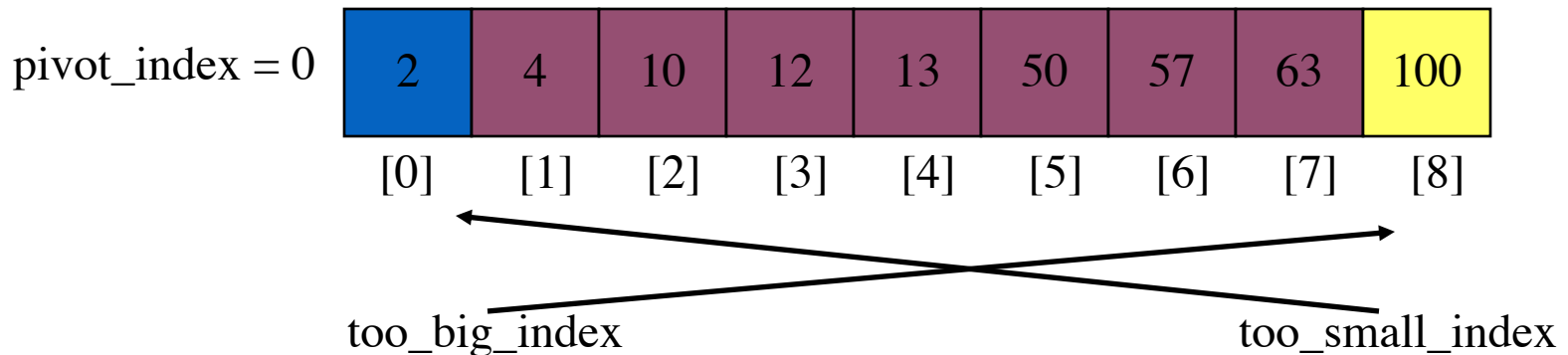
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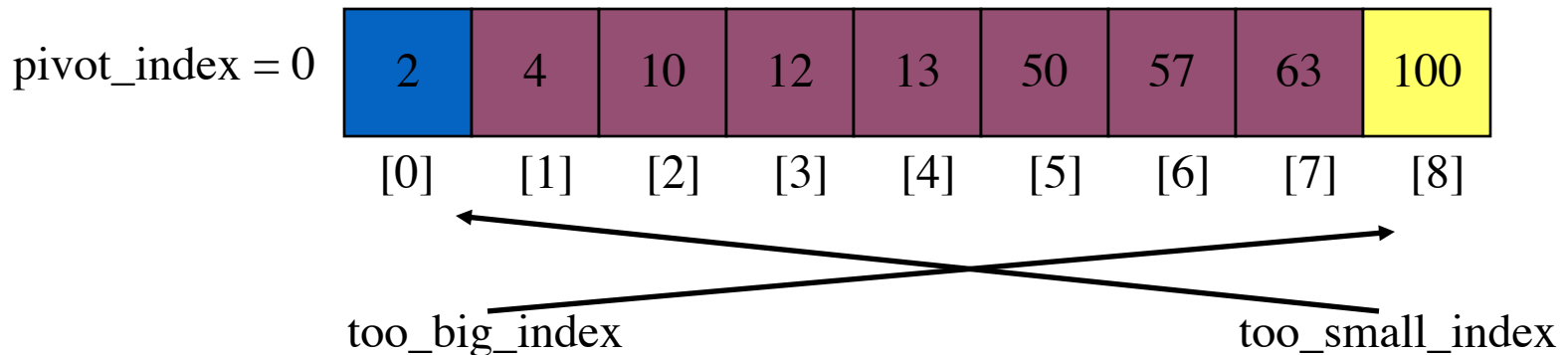
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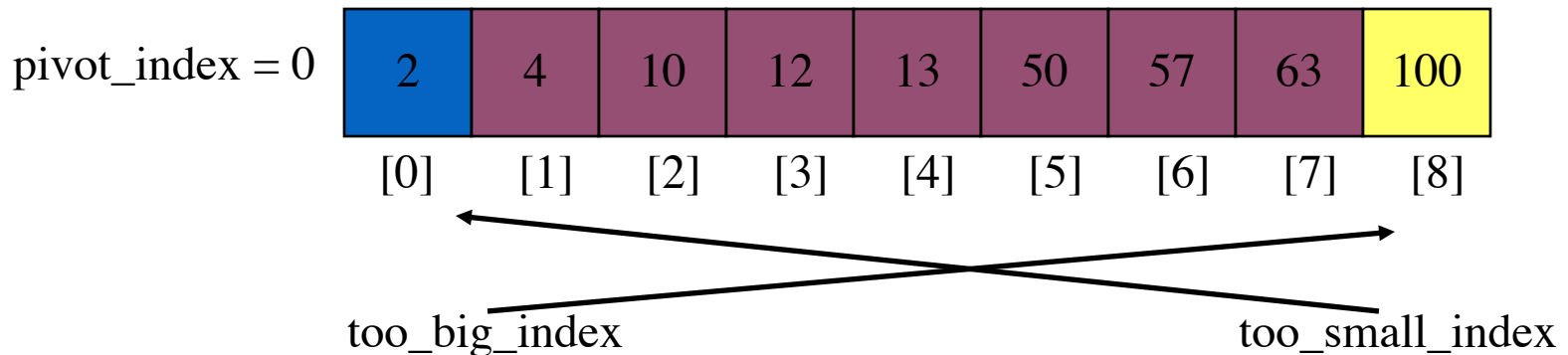
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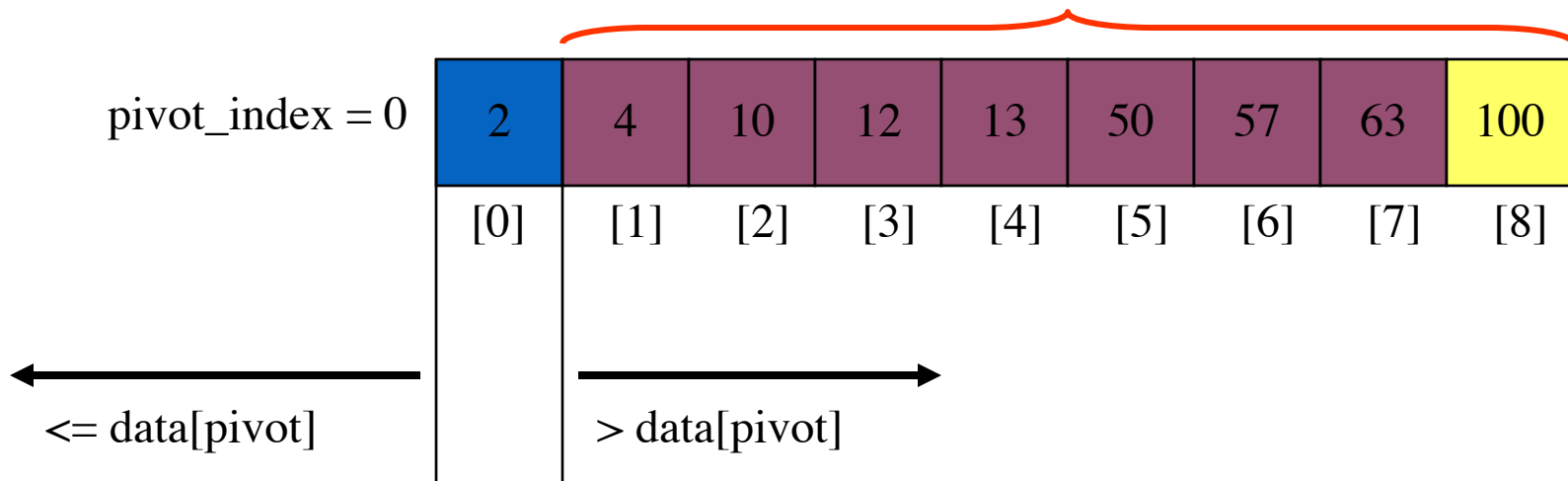
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Worst-Case Operation: Example

1. **while** $\text{data}[\text{too_big_index}] \leq \text{data}[\text{pivot}]$
 $\text{too_big_index} = \text{too_big_index} + 1$
2. **while** $\text{data}[\text{too_small_index}] > \text{data}[\text{pivot}]$
 $\text{too_small_index} = \text{too_small_index} - 1$
3. **if** $\text{too_big_index} < \text{too_small_index}$
 swap $\text{data}[\text{too_big_index}]$ and $\text{data}[\text{too_small_index}]$
4. **while** $\text{too_small_index} > \text{too_big_index}$, go to 1.
- 5. swap $\text{data}[\text{too_small_index}]$ and $\text{data}[\text{pivot_index}]$



Quicksort Analysis

- Assume that keys are random, uniformly distributed.
- Best case running time: $O(n \log_2 n)$

Quicksort Analysis

- Worst case running time?
- Recursion:
 - Partition splits array in two sub-arrays:
 - one sub-array of **size 0**
 - the other sub-array of **size n-1**
 - Quicksort each sub-array
- Depth of recursion tree?
 - **$O(n)$**
- Number of accesses per partition?
 - **$O(n)$**

Quicksort Analysis

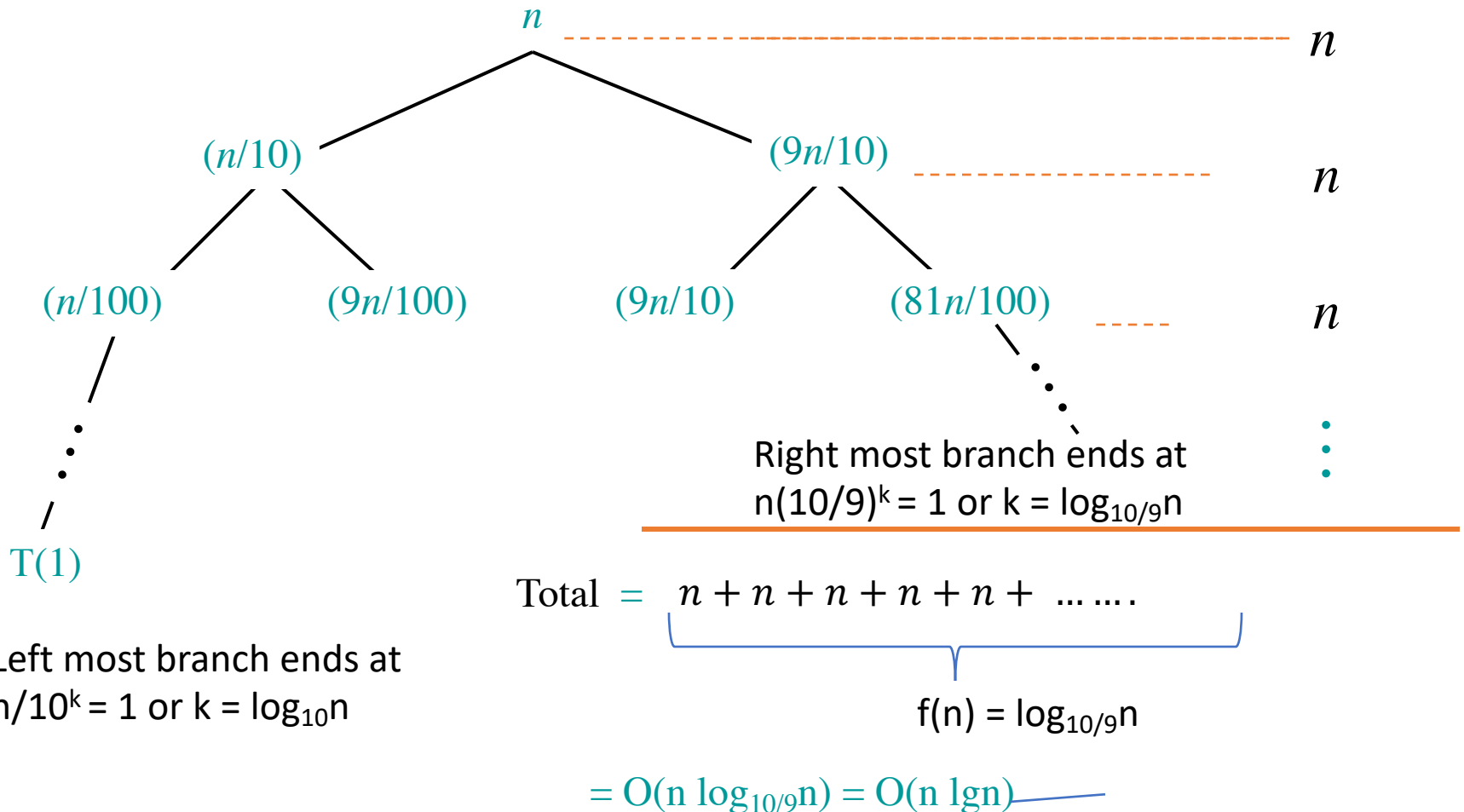
- **Best case running time:**
 $O(n \lg n)$
- **Worst case running time:**
 $O(n^2)$

QuickSort Analysis

- The performance of QuickSort depends on the size of the partitions
- Lop-sided partitions perform worse than balanced partitions
- However, **not all lop-sided partitions are bad**

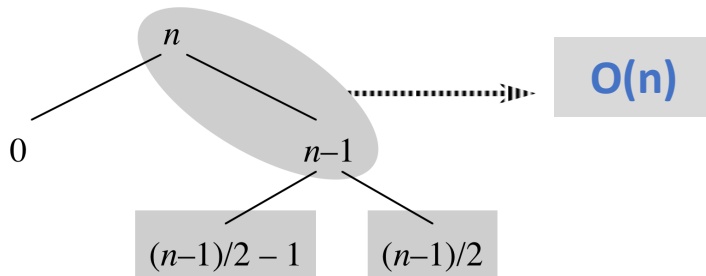
Remember this from Quiz 2?

Essentially QuickSort's performance when it produces a 9-1 split every time

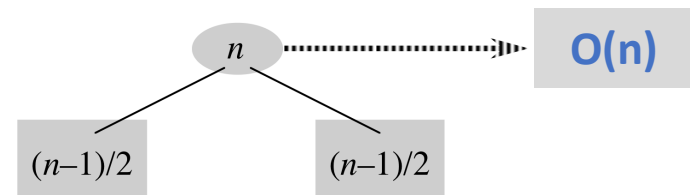


QuickSort Analysis

- It is the repeated **lop-sided partitioning** of of an list into sub-lists of size 0 and size $n-1$ that leads to $O(n^2)$ performance.
- What happens in the **average case**?
 - That is we alternate between one level with lop-sided partition and then next level we have balanced (back and forth) partition?



Cost of reaching here is $O(n) + O(n-1) \sim O(n)$



Cost of reaching here is $O(n)$

The cost of bad splits absorbed by the good splits

Improved Pivot Selection

- Notice, unlike InsertionSort, **QuickSort performs badly when the list is already sorted**
 - **Same holds true for reverse sorted array!!**
- What can we do to avoid worst case?
 - Pick **median value of three elements** from data array: $\text{data}[0]$, $\text{data}[n/2]$, and $\text{data}[n-1]$.
 - Use this median value as pivot.
 - Randomized QuickSort
 - You will play with this in the lab

QuickSort: Summary

- Quick sort:
 - Divide-and-conquer:
 - Partition array into two sub-arrays, recursively sort
 - All of first sub-array < all of second sub-array
 - Pro's:
 - $O(n \lg n)$ average case
 - Sorts in place
 - Fast in practice (*why?*)
 - Con's:
 - $O(n^2)$ worst case
 - Naïve implementation: worst case on sorted input
 - Good partitioning makes this very unlikely.

Non-Comparison Based Sorting

- Many times we have restrictions on our keys
 - Deck of cards: Ace->King and four suites
 - Social Security Numbers
 - Employee ID's
- We will examine an algorithm which under certain conditions can run in $O(n)$ time.
 - Counting sort
 - **Bucket Sort** (you will play with this in assignment 1)

Counting Sort

- Depends on assumption about the numbers being sorted
 - Assume numbers are in the range $1..k$
- The algorithm:
 - Input: $A[1..n]$, where $A[j] \in \{1, 2, 3, \dots, k\}$
 - Output: $B[1..n]$, sorted (not sorted in place)
 - Also: Array $C[1..k]$ for auxiliary storage
- Therefore needs $O(|B| + |C|)$ extra storage
 - Which is same as $O(n+k)$

Counting Sort Pseudocode)

Input list

Size of A (and B)

Range of numbers

CountingSort(A, n, k)

B = [], C = []

for i=0 to k

C[i] = 0

for j=0 to n

C[A[j]] += 1

for i=1 to k

C[i] = C[i] + C[i-1]

for j=n-1 *downto* 0

B[C[A[j]]-1] = A[j]

C[A[j]] -= 1

return B

This is called
a *histogram*.

Example

A

2	5	3	0	2	3	0	3
---	---	---	---	---	---	---	---

C

2	0	2	3	0	1
---	---	---	---	---	---

After **first** for loop

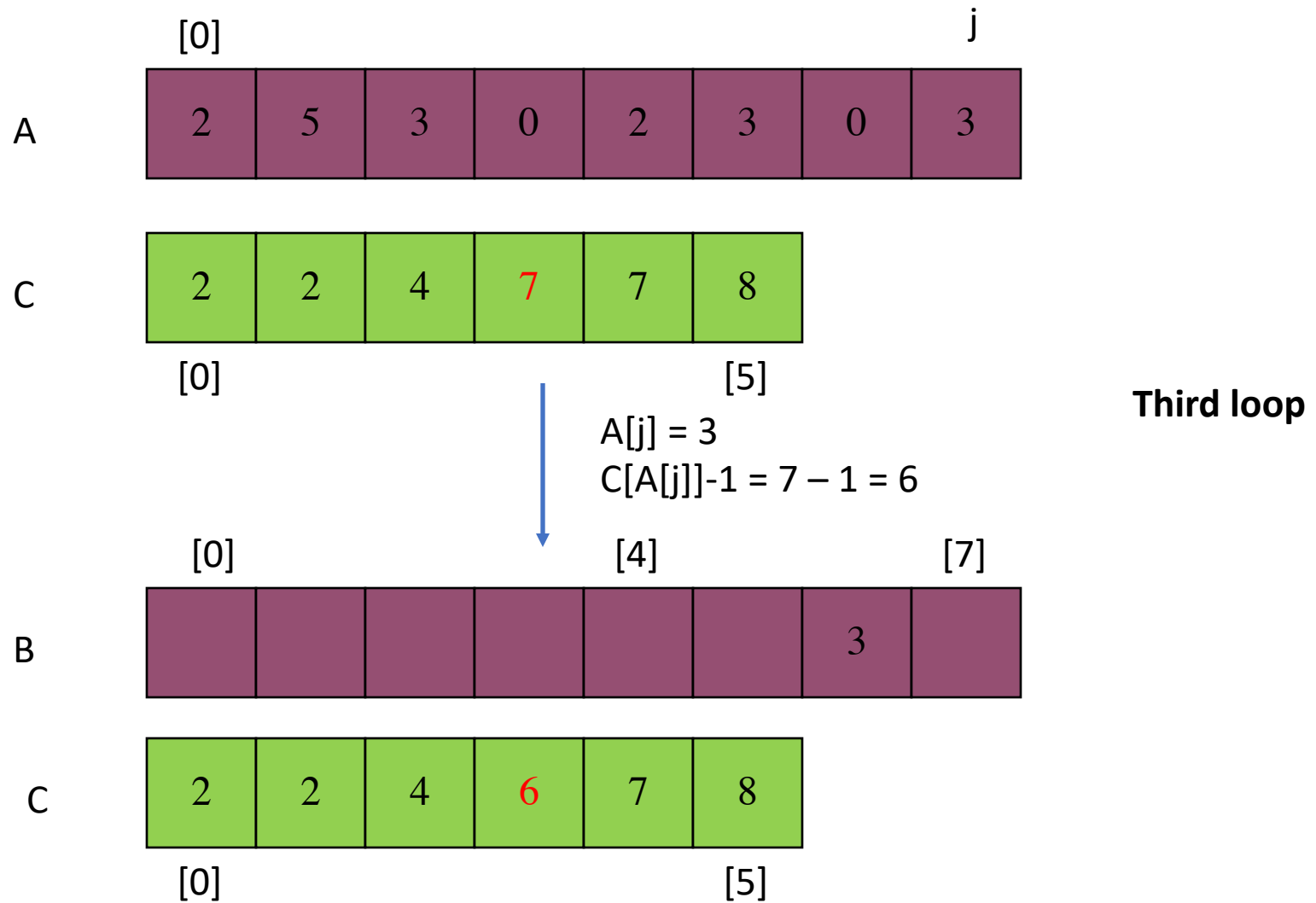
C

2	2	4	7	7	8
---	---	---	---	---	---

After **second** for loop

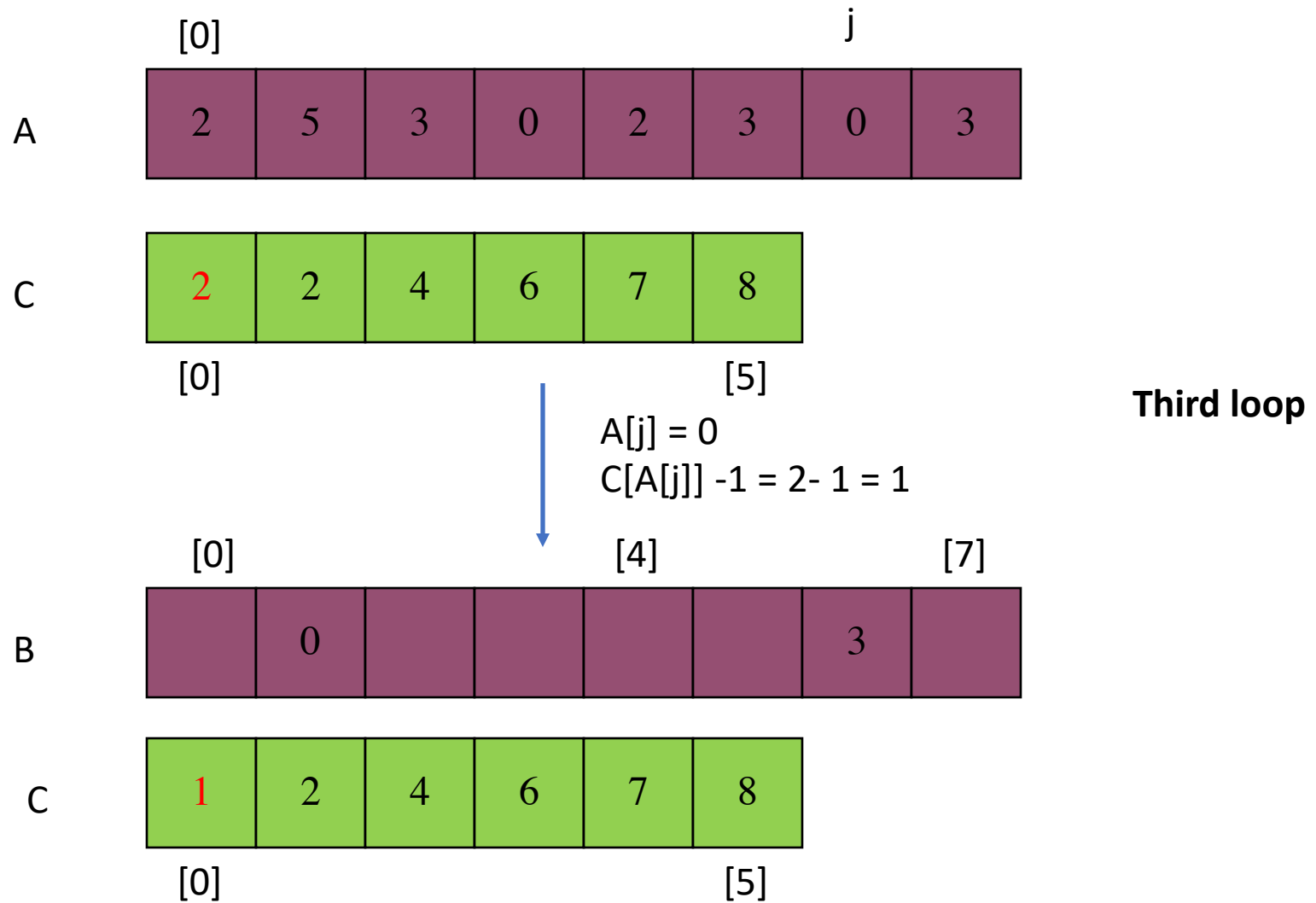
k = range of numbers = (0-5)

Example



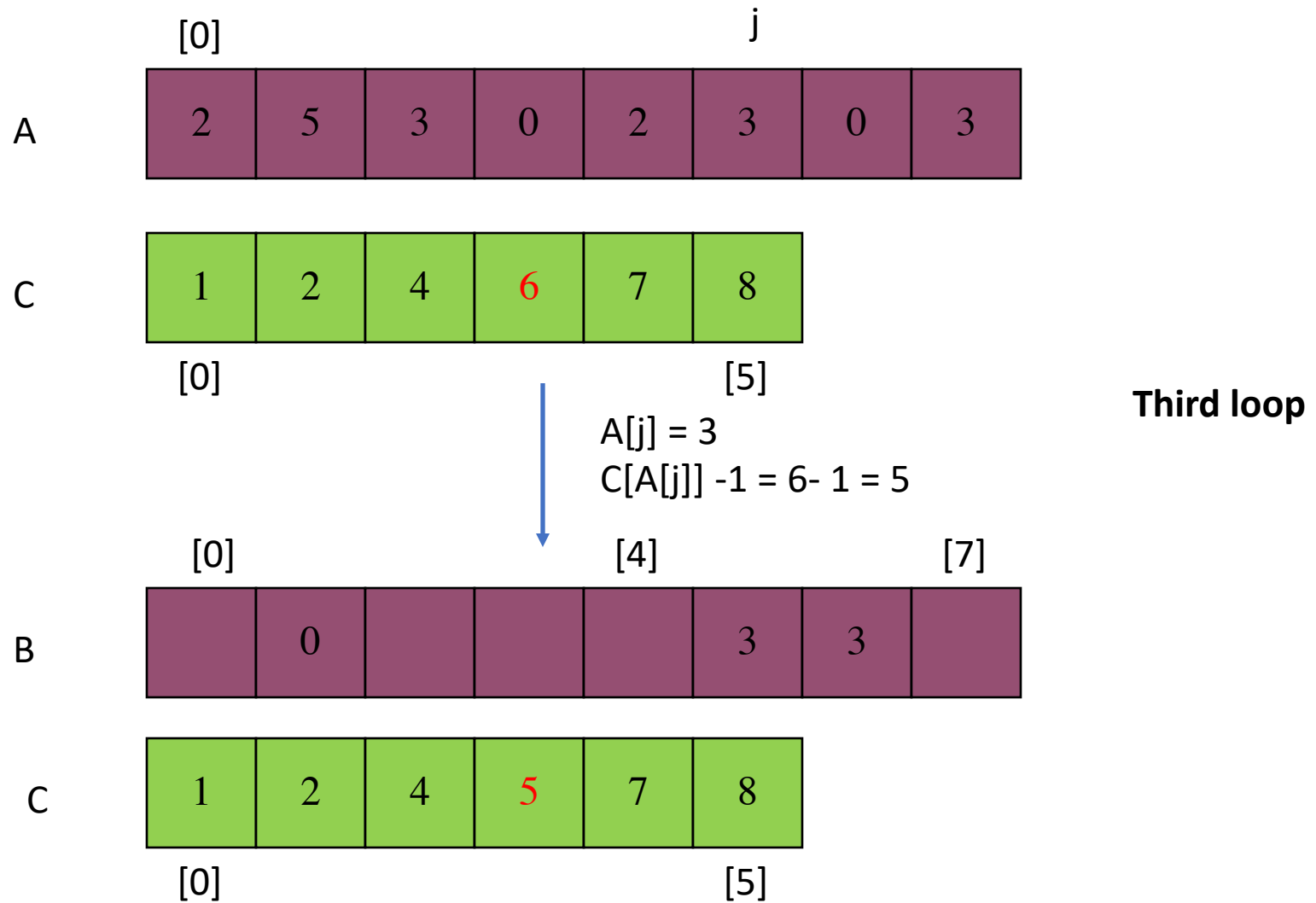
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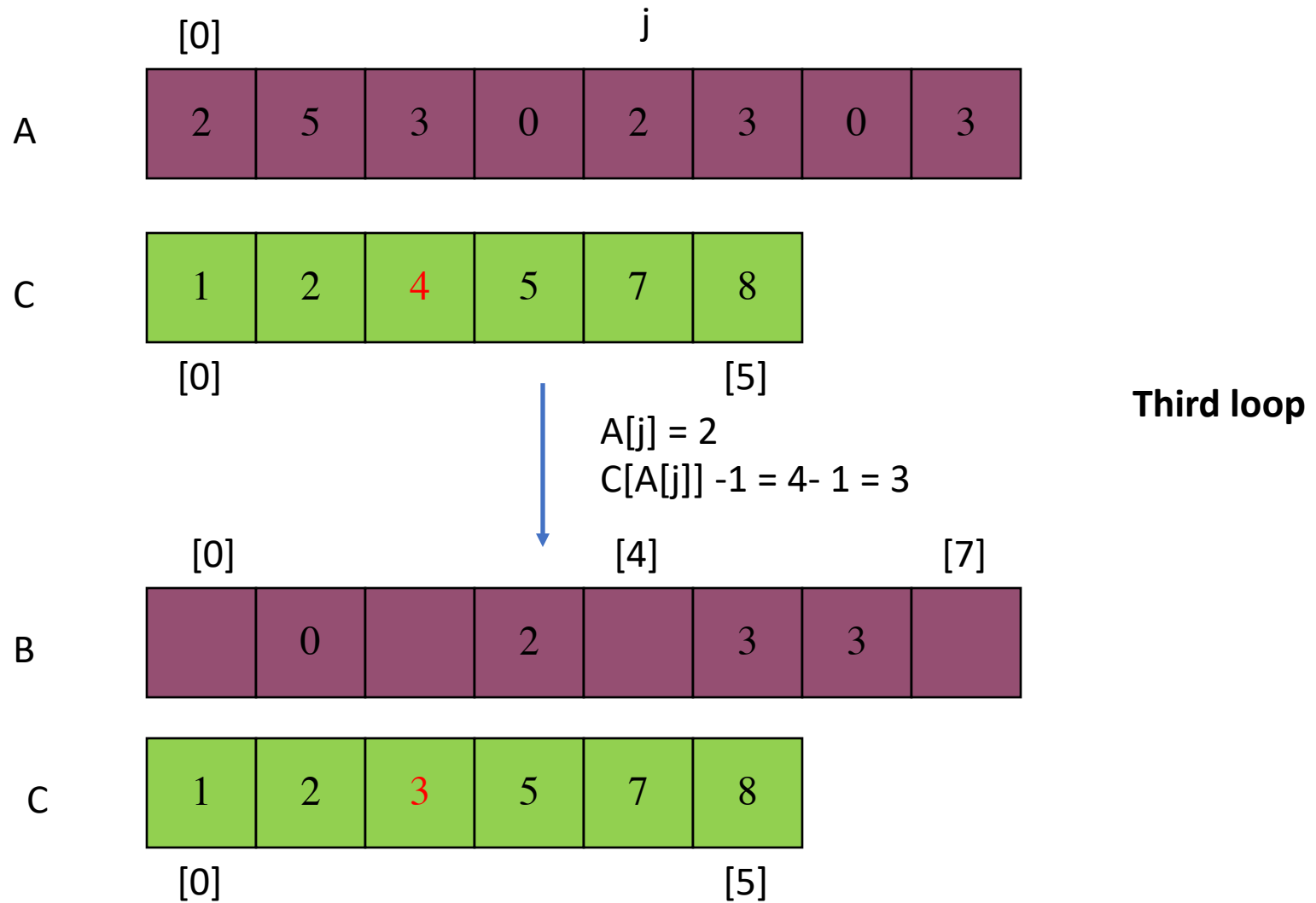
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Example



$k = \text{range of numbers} = (0-5)$

Example



$k = \text{range of numbers} = (0-5)$

Example

	[0]			j				
A	2	5	3	0	2	3	0	3

C	1	2	3	5	7	8
---	---	---	---	---	---	---

[0] [5]

$$A[j] = 0$$
$$C[A[j]] - 1 = 1 - 1 = 0$$

Third loop

	[0]			[4]			[7]
B	0	0		2		3	3

C	0	2	3	5	7	8
---	---	---	---	---	---	---

[0] [5]

k = range of numbers = (0-5)

Example

A

[0]			j				
2	5	3	0	2	3	0	3

C

0	2	3	5	7	8
---	---	---	---	---	---

[0] [5]

$A[j] = 3$
 $C[A[j]] - 1 = 5 - 1 = 4$

Third loop

B

[0]				[4]			[7]
0	0		2	3	3	3	

C

0	2	3	4	7	8
---	---	---	---	---	---

[0] [5]

k = range of numbers = (0-5)

Example

	[0]	j						
A	2	5	3	0	2	3	0	3

C	0	2	3	4	7	8
---	---	---	---	---	---	---

[0]

[5]

$A[j] = 5$

$C[A[j]] - 1 = 8 - 1 = 7$

Third loop

[0]

[4]

[7]

B	0	0		2	3	3	3	5
---	---	---	--	---	---	---	---	---

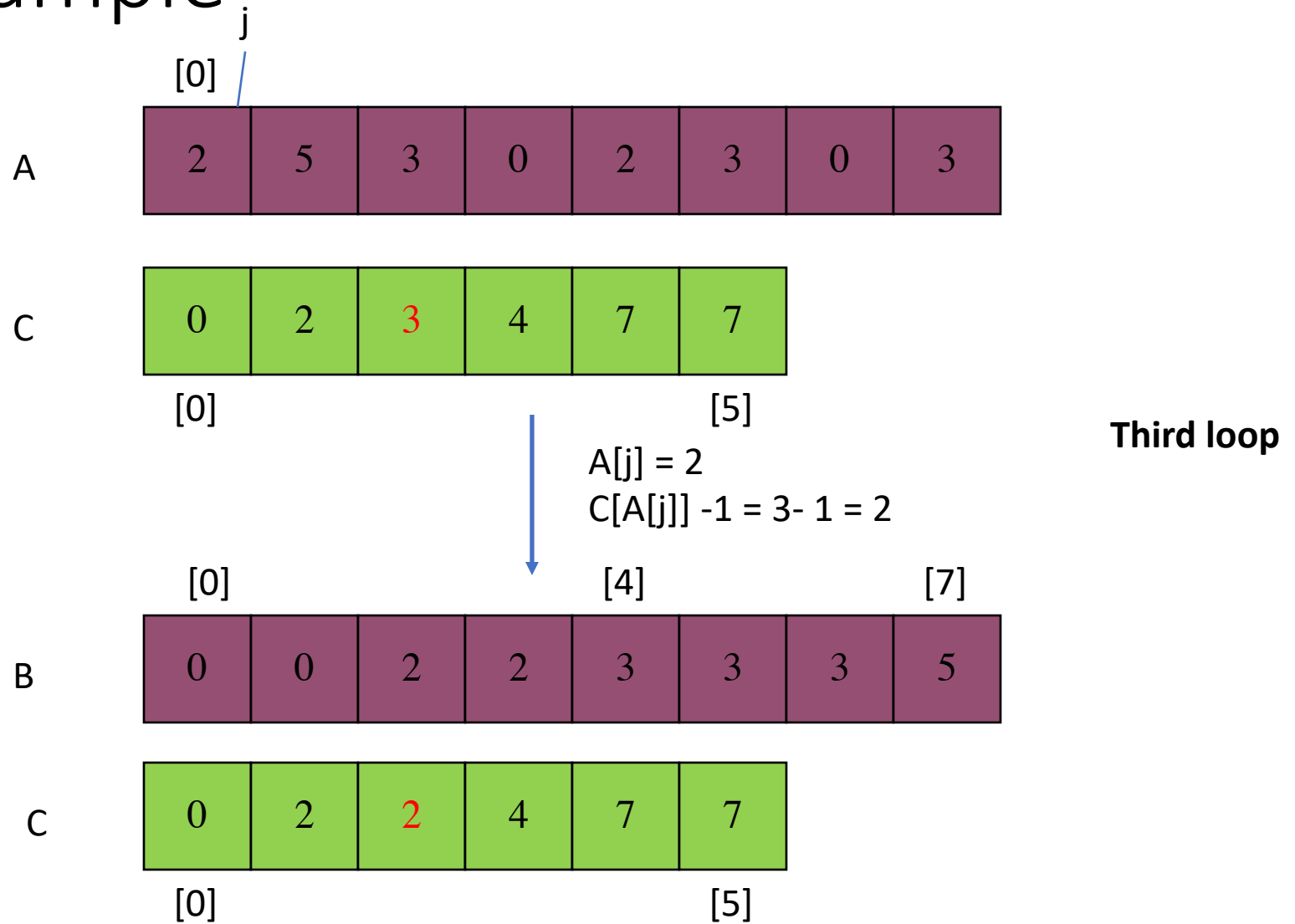
C	0	2	3	4	7	7
---	---	---	---	---	---	---

[0]

[5]

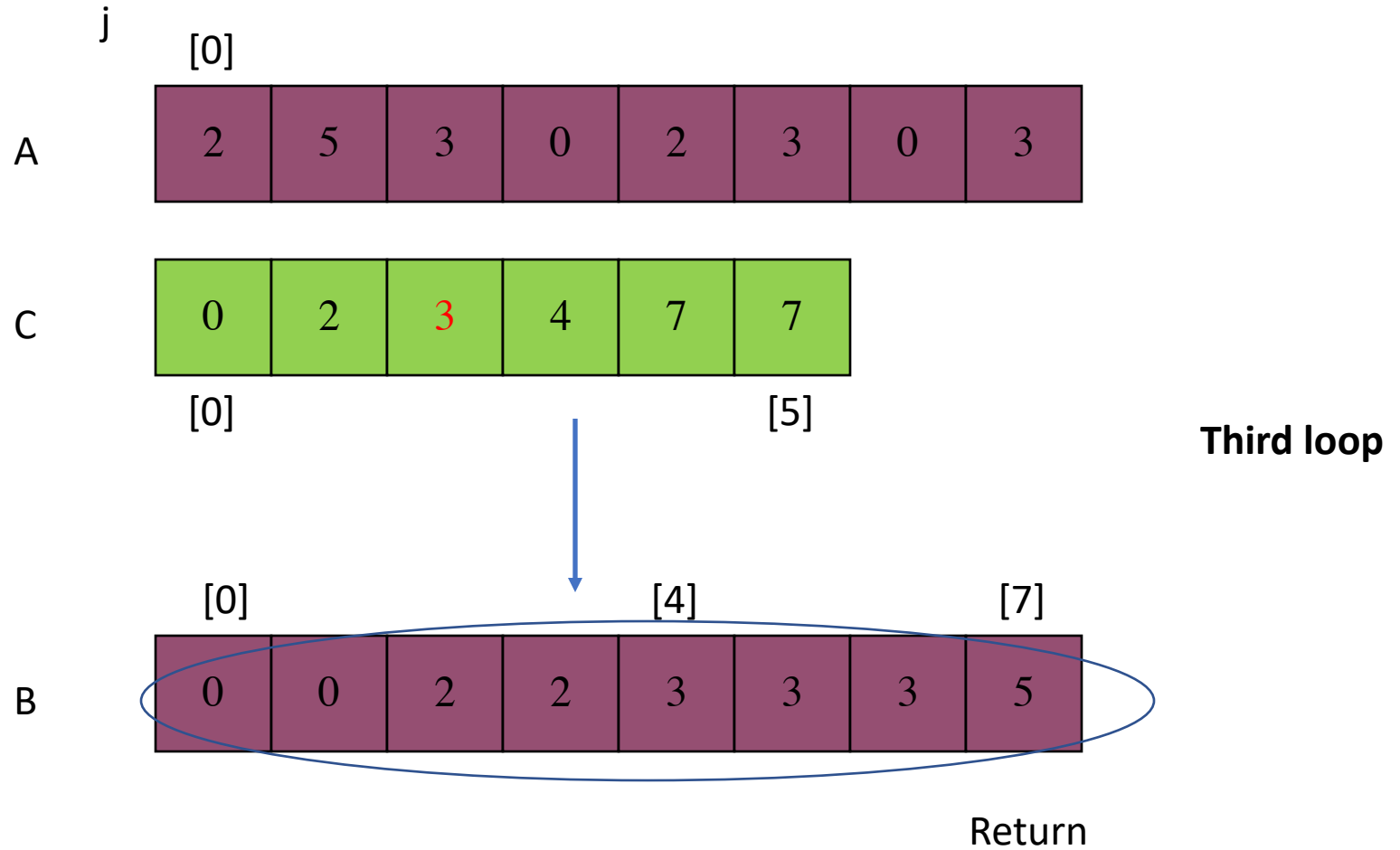
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Example



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Example



k = range of numbers = (0-5)

Counting Sort

```
1  CountingSort(A, B, k)
2      for i=1 to k
3          C[i] = 0;
4      for j=1 to n
5          C[A[j]] += 1;
6      for i=2 to k
7          C[i] = C[i] + C[i-1];
8      for j=n downto 1
9          B[C[A[j]]] = A[j];
10         C[A[j]] -= 1;
```

Takes time $O(k)$

Takes time $O(n)$

What is the running time?

Counting Sort

- Total time: $O(n + k)$
 - Works well if $k = O(n)$ or $k = O(1)$
- This sorting is *stable*.
 - A sorting algorithm is ***stable*** when numbers with the same values appear in the output array in the same order as they do in the input array.

Counting Sort: Summary

- **Assumption:** input taken from **small** set of **numbers** of size k
- Basic idea:
 - Count number of elements less than you for each element.
 - This gives the position of that number – similar to selection sort.
- Pro's:
 - Fast
 - Asymptotically fast - $O(n+k)$
 - Simple to code
- Con's:
 - Doesn't sort in place.
 - Elements must be integers.
 - Requires $O(n+k)$ extra storage.



That's all Folks!
Any Question?