

CS4102 Algorithms

Fall 2019

Warm up

Show that finding the minimum of an unordered list requires $\Omega(n)$ comparisons

Find Min, Lower Bound Proof

Show that finding the minimum of an unordered list requires $\Omega(n)$ comparisons

Suppose (toward contradiction) that there is an algorithm for Find Min that does fewer than $\frac{n}{2} = \Omega(n)$ comparisons.

This means there is at least one “uncompared” element
We can't know that this element wasn't the min!

2	8	19	20		3	9	-4
0	1	2	3	4	5	6	7

Homeworks

- HW3 due 11pm Tuesday, October 1
 - Divide and conquer
 - Written (use LaTeX!)
 - Submit **BOTH** a pdf and a zip file (2 separate attachments)
- Regrade Office Hours
 - Thursdays 11am-12pm @ Rice 210 (starting next week!)
 - Thursdays 4pm-5pm @ Rice 501 (starting today!)

Today's Keywords

- Sorting
- Linear time Sorting
- Counting Sort
- Radix Sort
- Maximum Sum Continuous Subarray

CLRS Readings

- Chapter 8

Sorting, so far

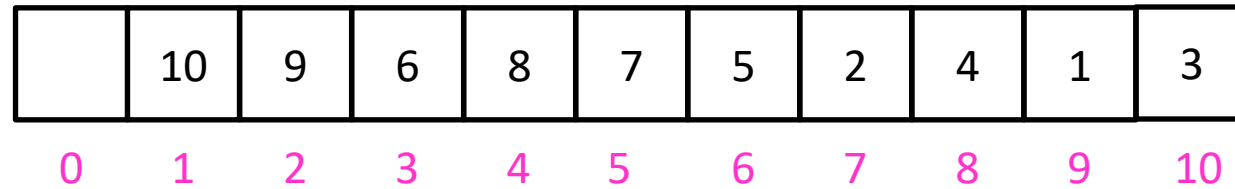
- Sorting algorithms we have discussed:
 - Mergesort $O(n \log n)$ Optimal!
 - Quicksort $O(n \log n)$ Optimal!
- Other sorting algorithms (will discuss):
 - Bubblesort $O(n^2)$
 - Insertionsort $O(n^2)$
 - Heapsort $O(n \log n)$ Optimal!

Speed Isn't Everything

- Important properties of sorting algorithms:
- **Run Time**
 - Asymptotic Complexity
 - Constants
- **In Place (or In-Situ)**
 - Done with only constant additional space
- **Adaptive**
 - Faster if list is nearly sorted
- **Stable**
 - Equal elements remain in original order
- **Parallelizable**
 - Runs faster with multiple computers

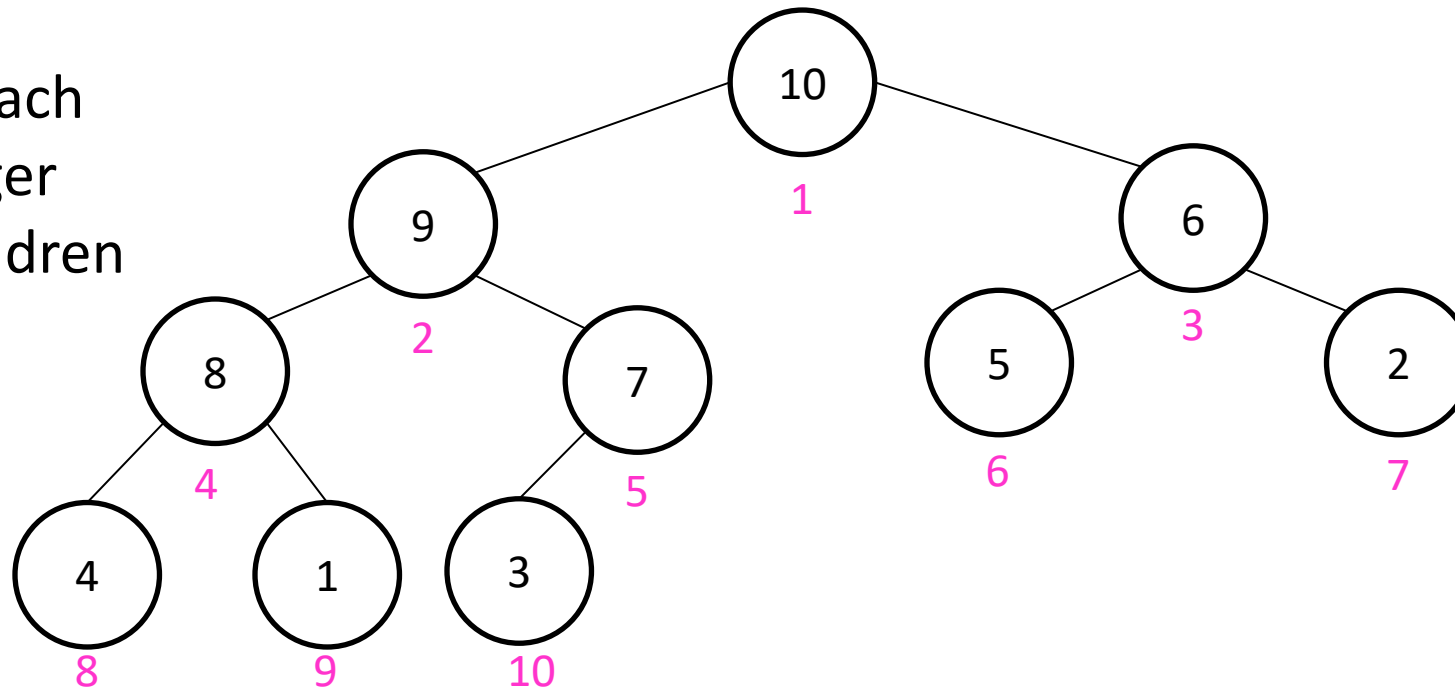
Heap Sort

- **Idea:** Build a Heap, repeatedly extract max element from the heap to build sorted list Right-to-Left



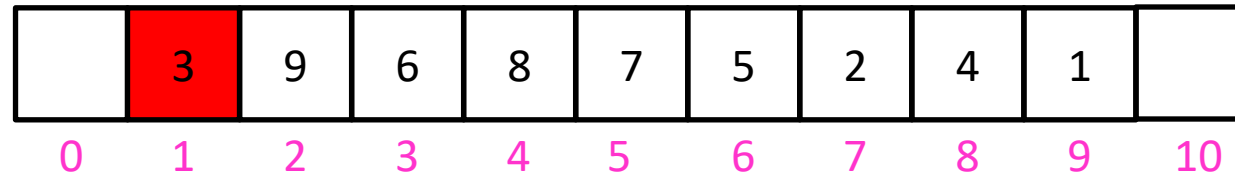
Max Heap

Property: Each node is larger than its children



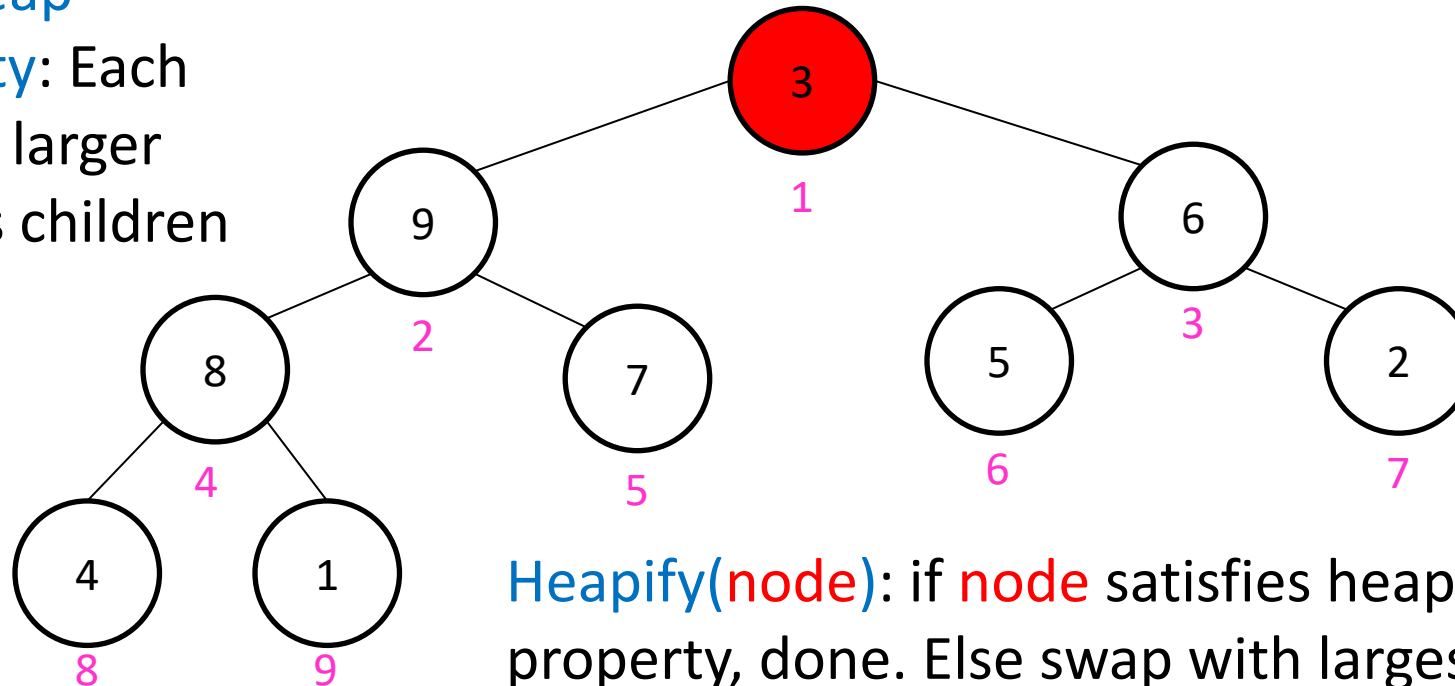
Heap Sort

- Remove the Max element (i.e. the root) from the Heap: replace with last element, call Heapify(root)



Max Heap

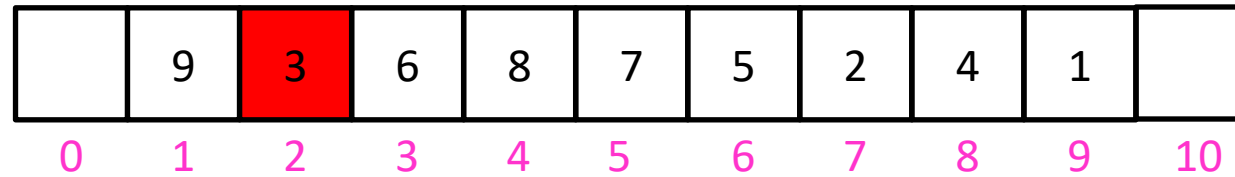
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Heapify(node): if node satisfies heap property, done. Else swap with largest child and recurse on that subtree

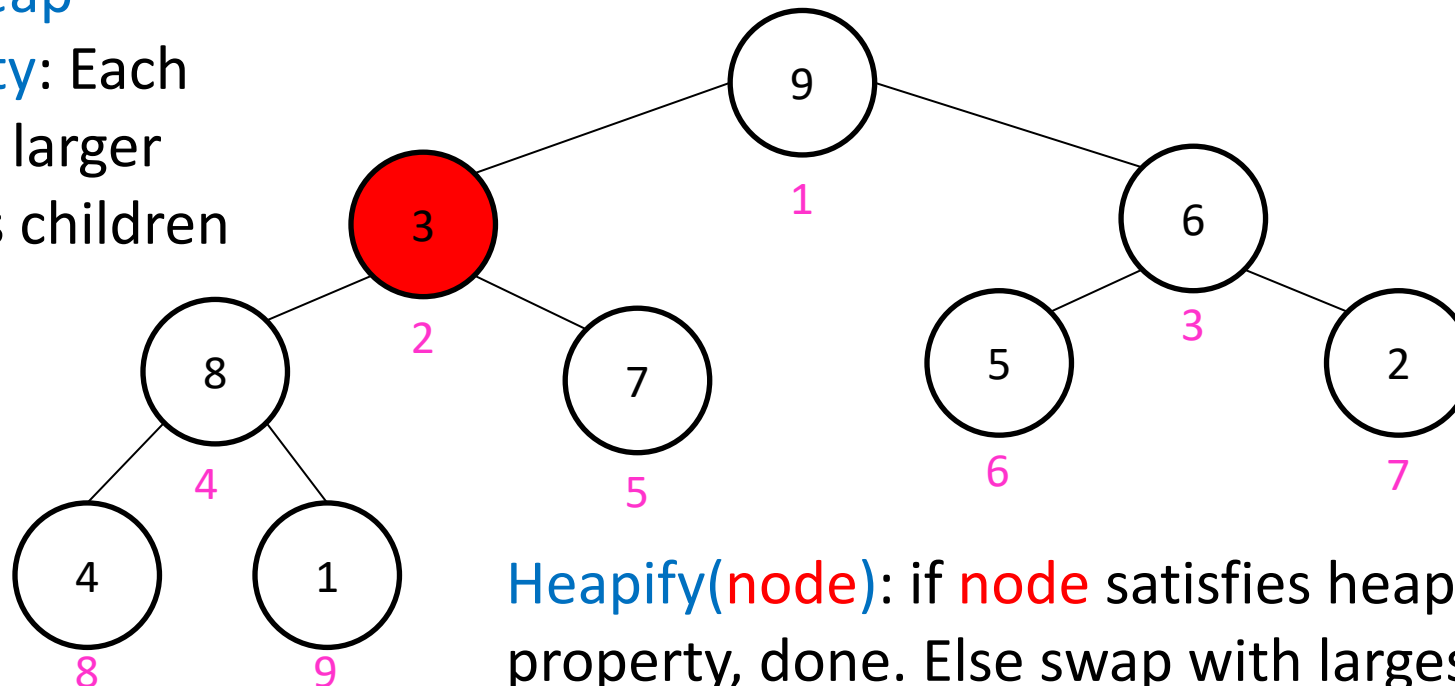
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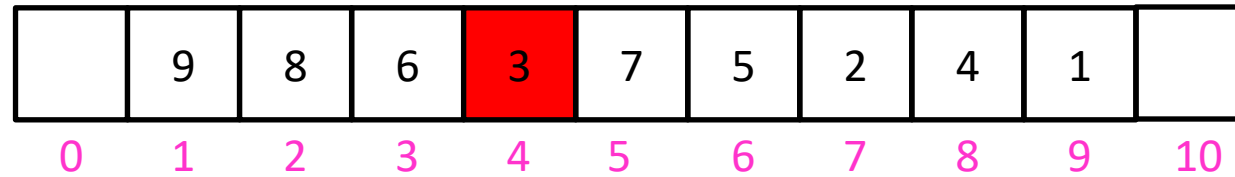
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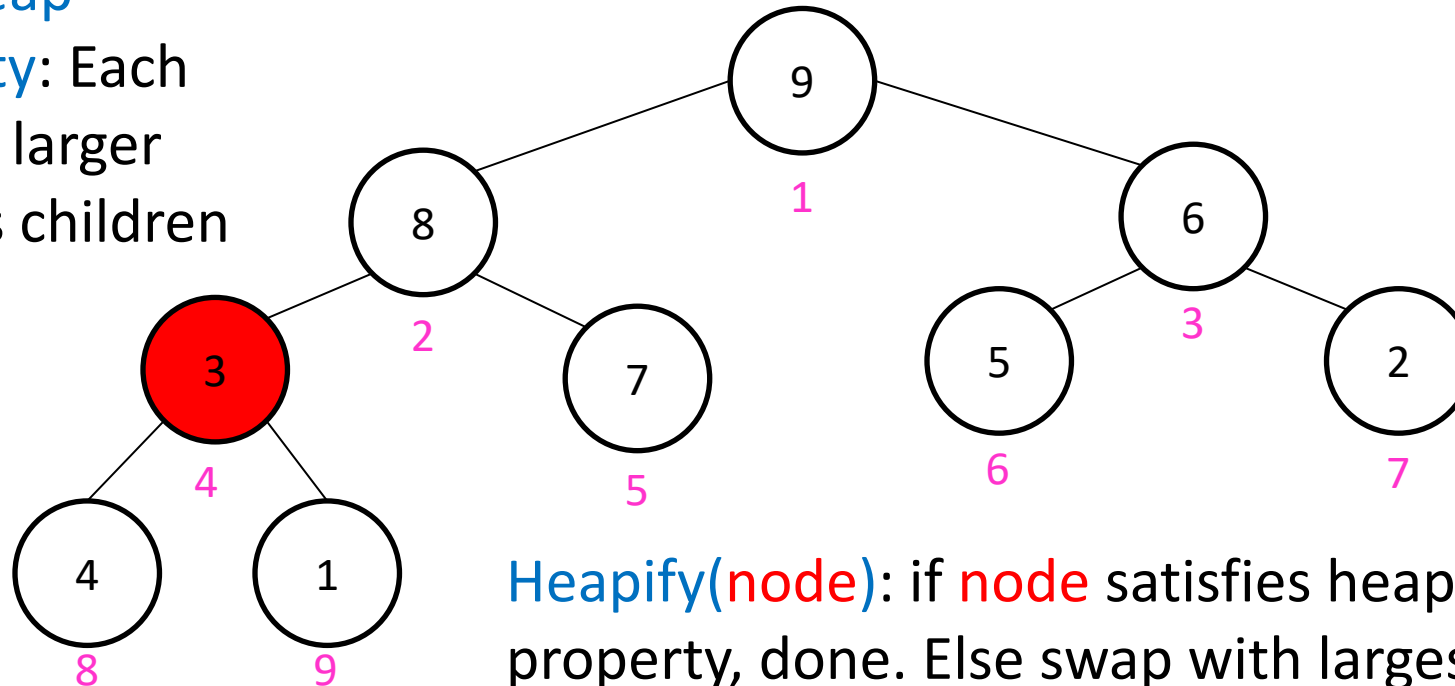
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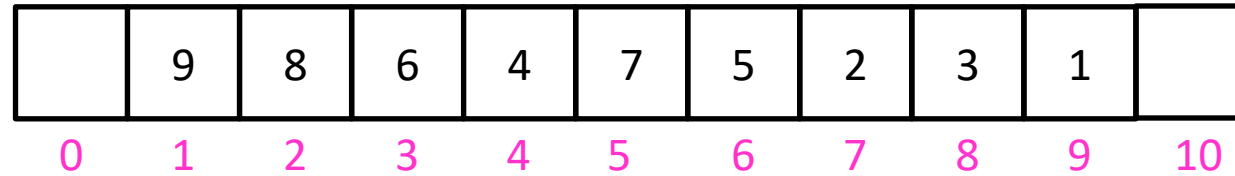
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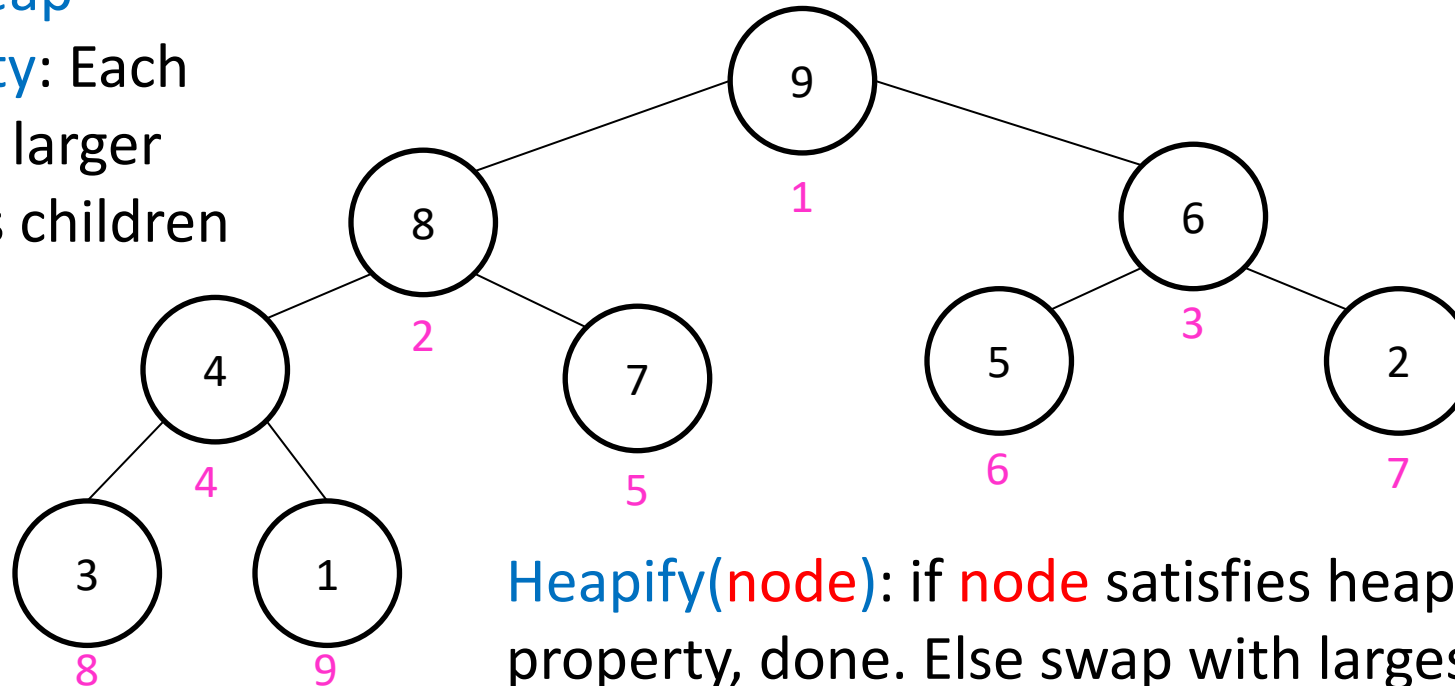
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In Place?

Yes!

When removing an element from the heap, move it to the (now unoccupied) end of the list

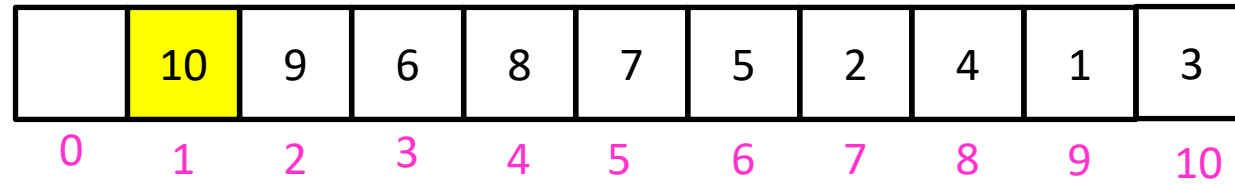
Run Time?

$\Theta(n \log n)$

Constants worse than Quick Sort

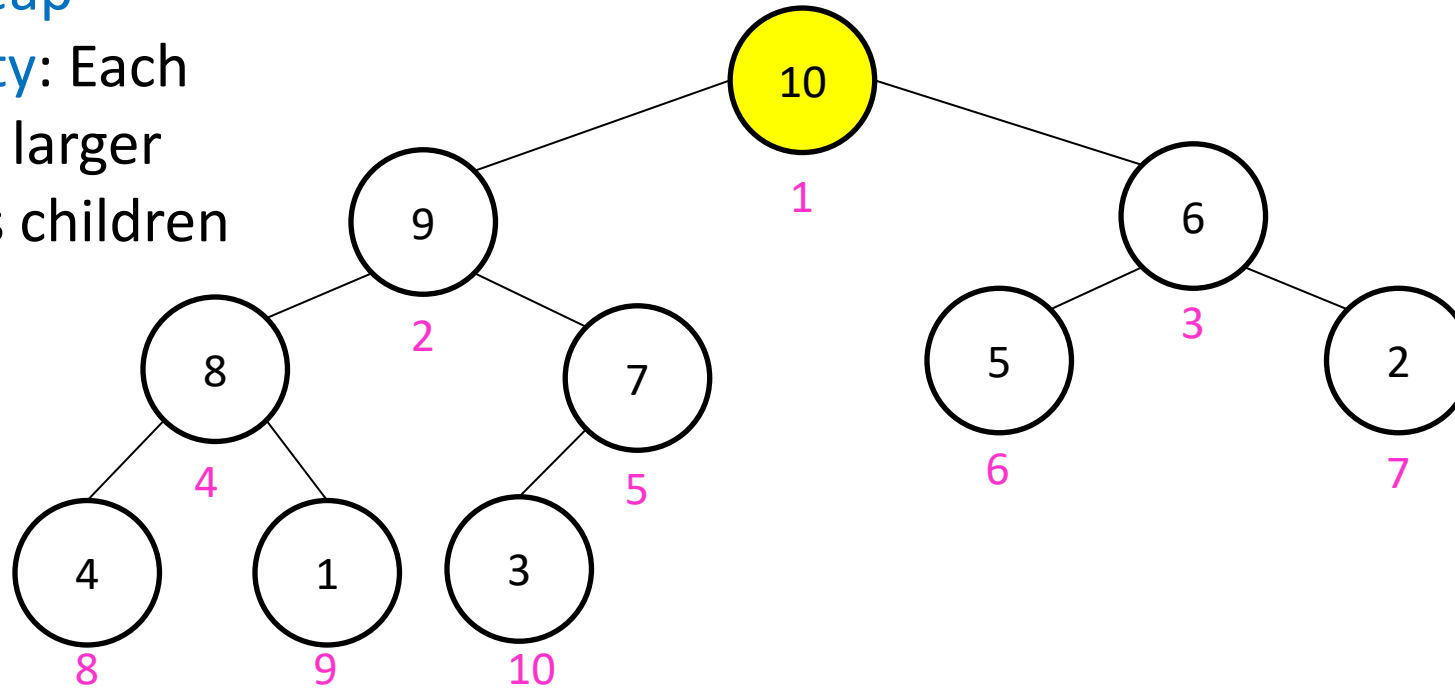
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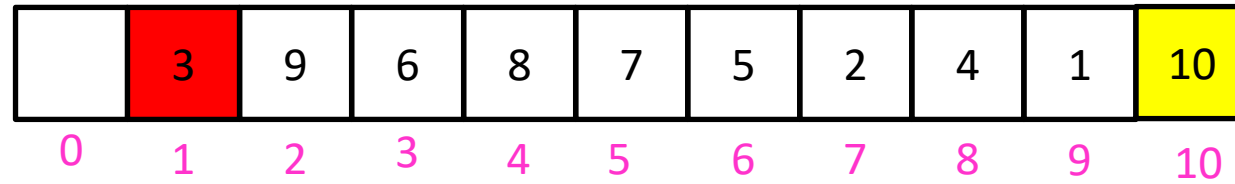
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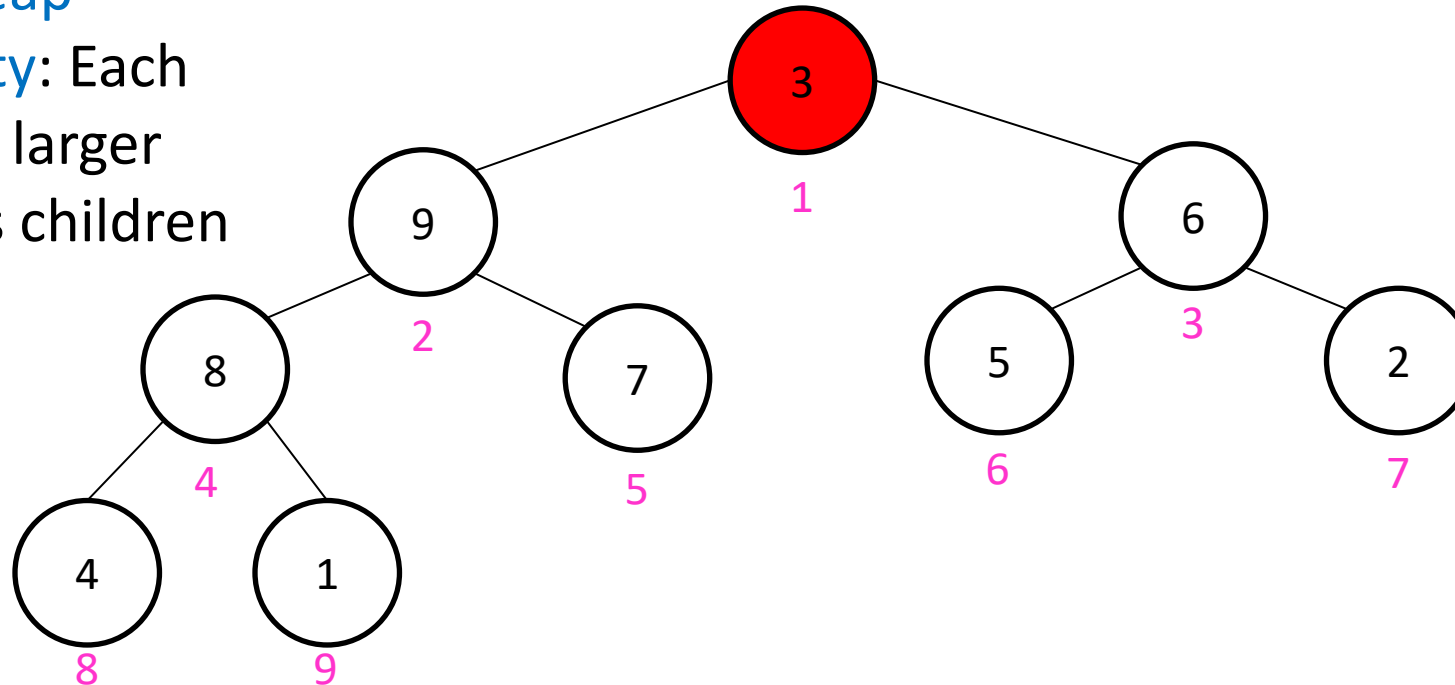
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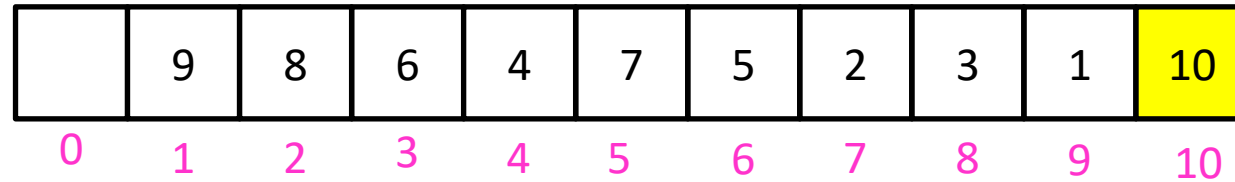
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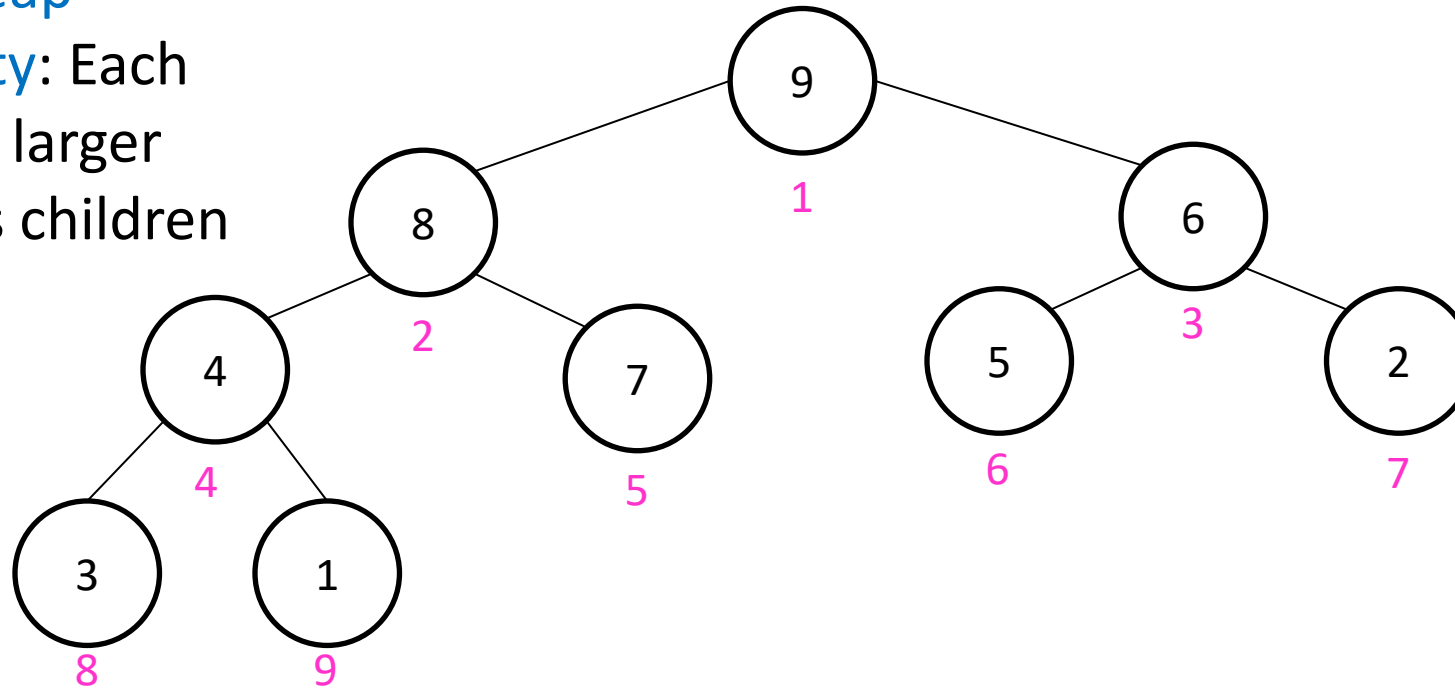
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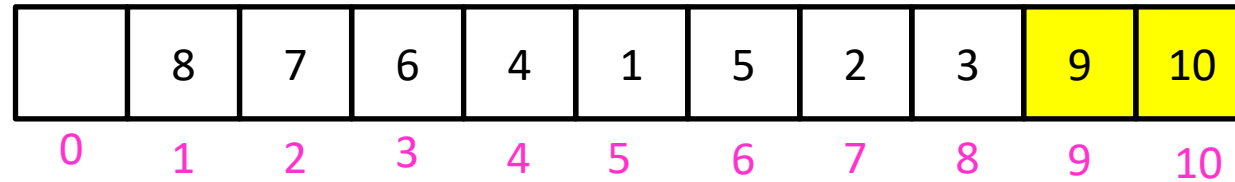
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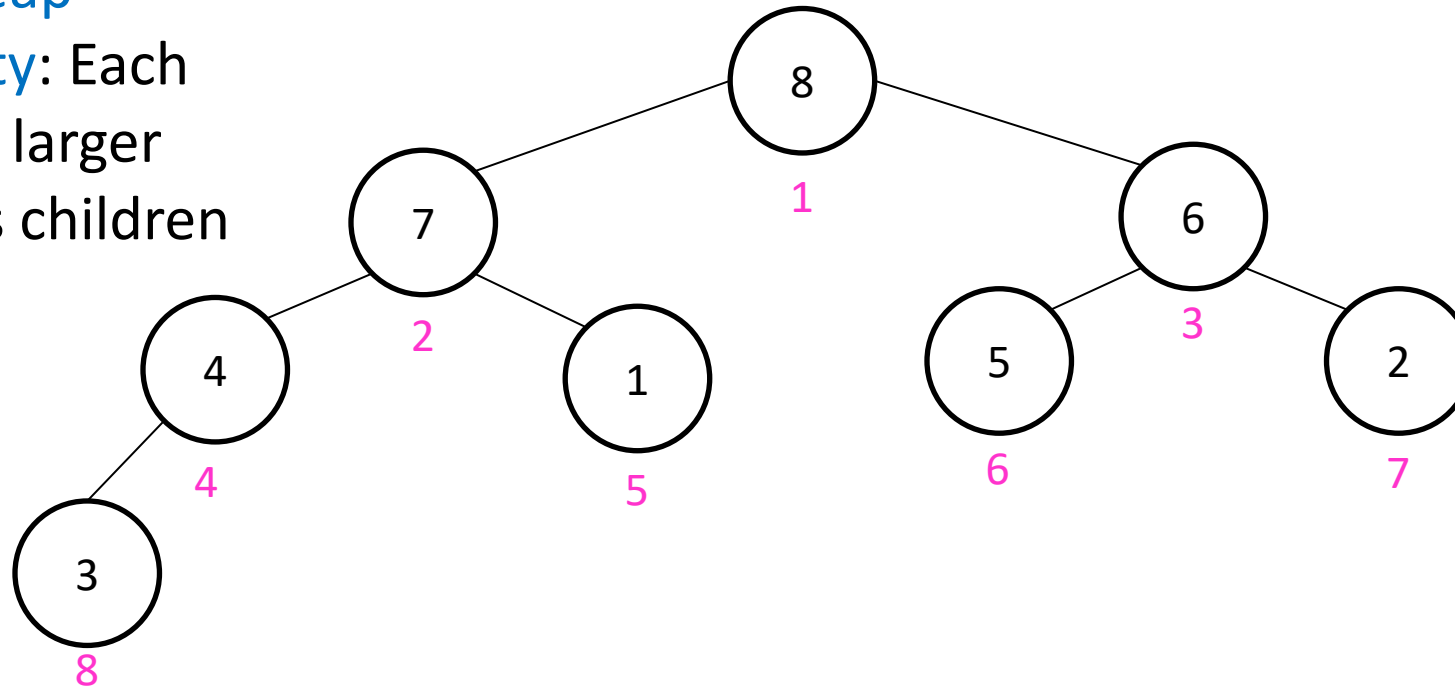
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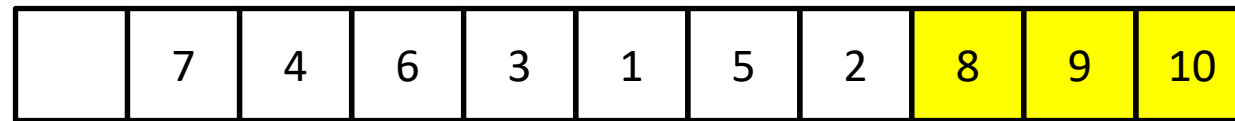
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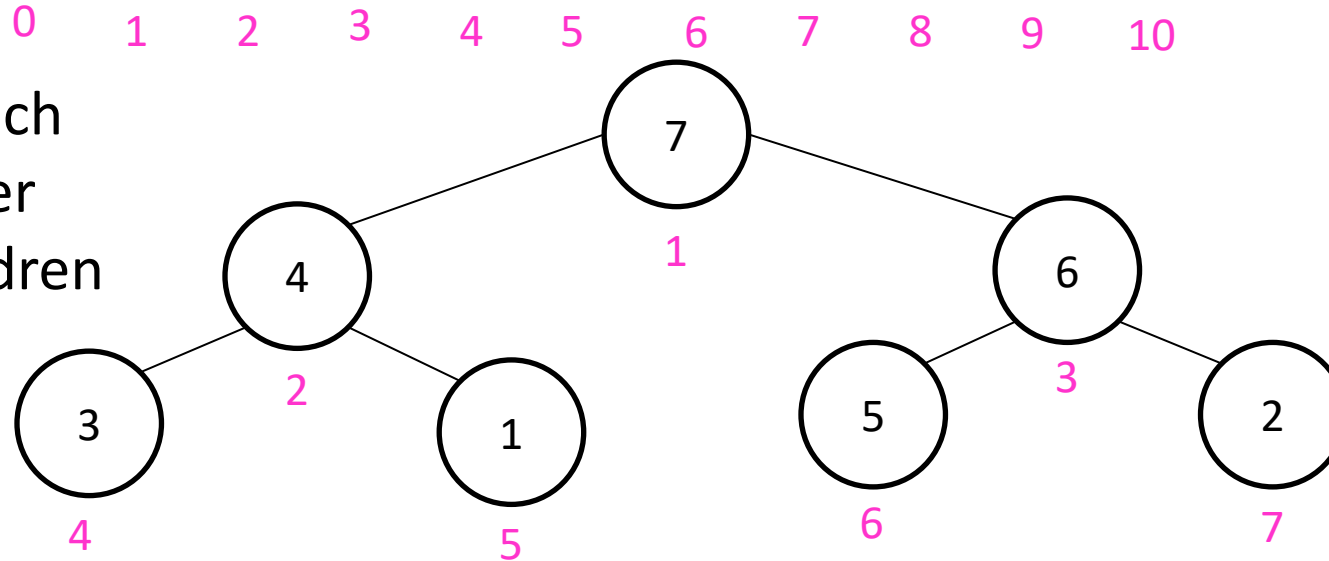
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Run Time?

$\Theta(n \log n)$

Constants worse than Quick Sort

Parallelizable?

In Place?

Yes!

Adaptive?

No

Stable?

No

No

Sorting, so far

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Sorting in Linear Time

- Cannot be comparison-based
- Need to make some sort of assumption about the contents of the list
 - Small number of unique values
 - Small range of values
 - Etc.

Counting Sort

- Idea: **Count** how many things are less than each element

$L =$

3	6	4	1	3	6	1	6
1	2	3	4	5	6	7	8

1. Range is $[1, k]$ (here $[1, 6]$)
make an array C of size k
populate with counts of each value

For i in L :
 $++C[L[i]]$

$C =$

2	0	2	1	0	3
1	2	3	4	5	6

running sum
↓

$C =$

2	2	4	5	5	8
1	2	3	4	5	6

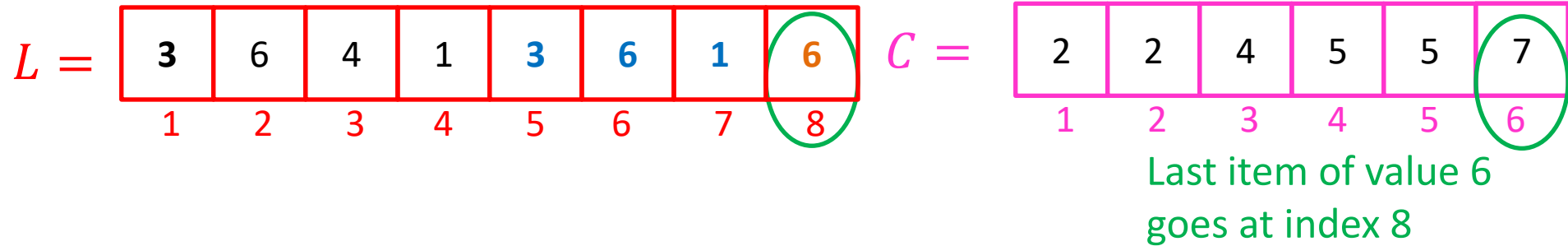
2. Take “running sum” of C
to count things less than each value

For $i = 1$ to $\text{len}(C)$:
 $C[i] = C[i - 1] + C[i]$

To sort: last item of
value 3 goes at index 4

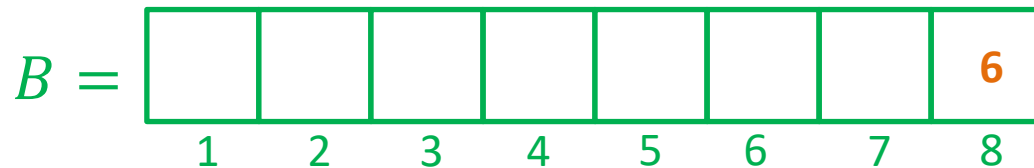
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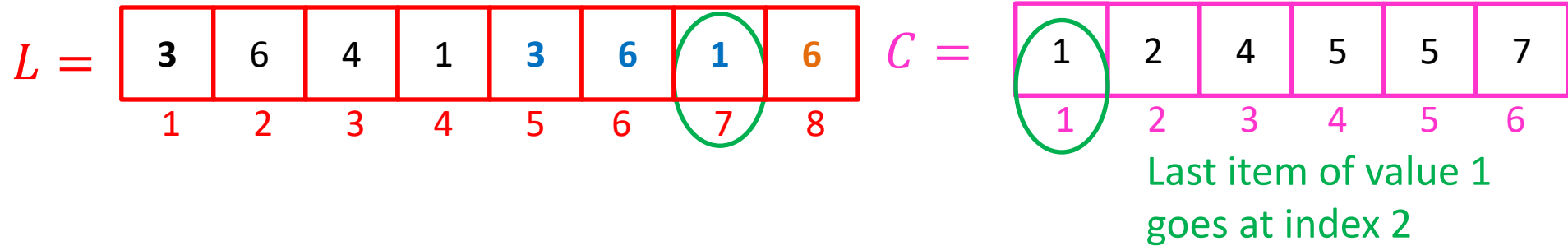
For each **element** of L (last to first):
Use C to find its **proper place in B**
Decrement that position of C

For $i = \text{len}(L)$ downto 1:
 $B[C[L[i]]] = L[i]$
 $C[L[i]] = C[L[i]] - 1$



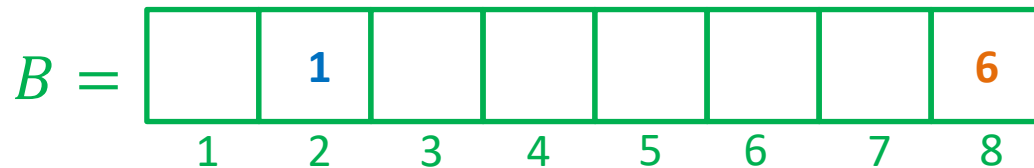
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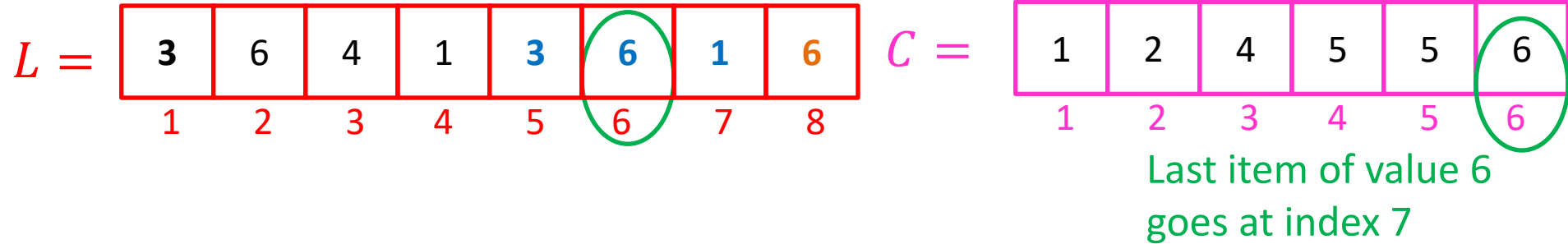
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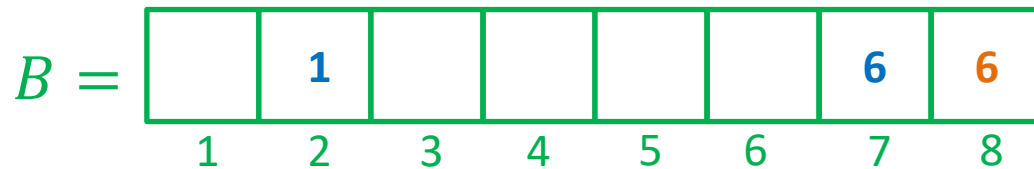
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Run Time: $O(n + k)$

Memory: $O(n + k)$

Counting Sort

- Why not always use counting sort?
- For 64-bit numbers, requires an array of length $2^{64} > 10^{19}$
 - 5 GHz CPU will require > 116 years to initialize the array
 - 18 Exabytes of data
 - Total amount of data that Google has

12 Exabytes



Radix Sort

- **Idea:** **Stable sort** on each digit, from least significant to most significant

103	801	401	323	255	823	999	101	113	901	555	512	245	800	018	121
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

Place each element into a “bucket” according to its 1’s place

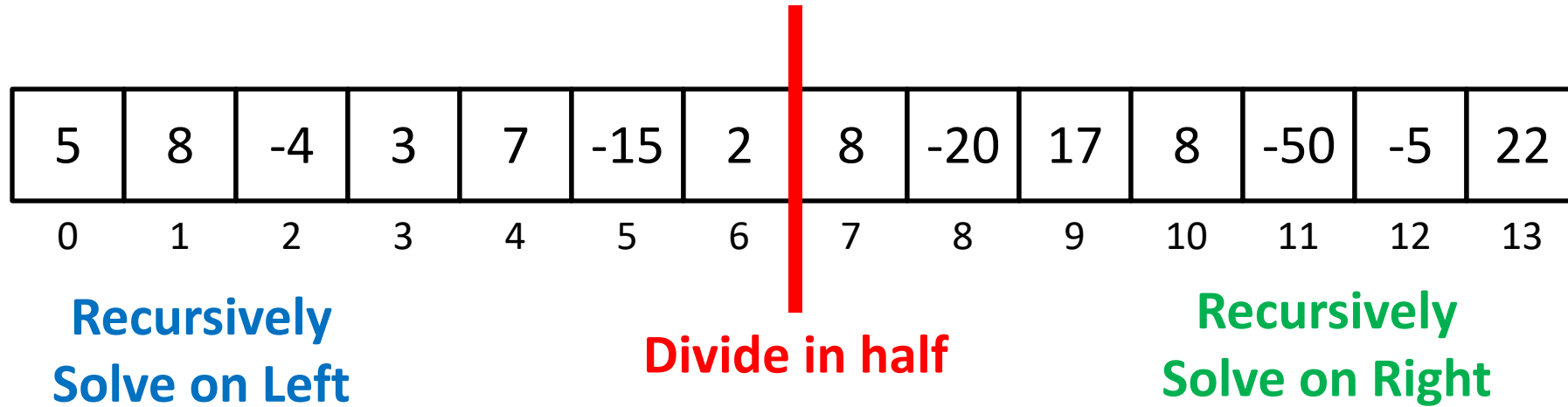
800	801 401 101 901 121	512	103 323 823 113		255 555 245			018	999
0	1	2	3	4	5	6	7	8	9

Maximum Sum Contiguous Subarray Problem

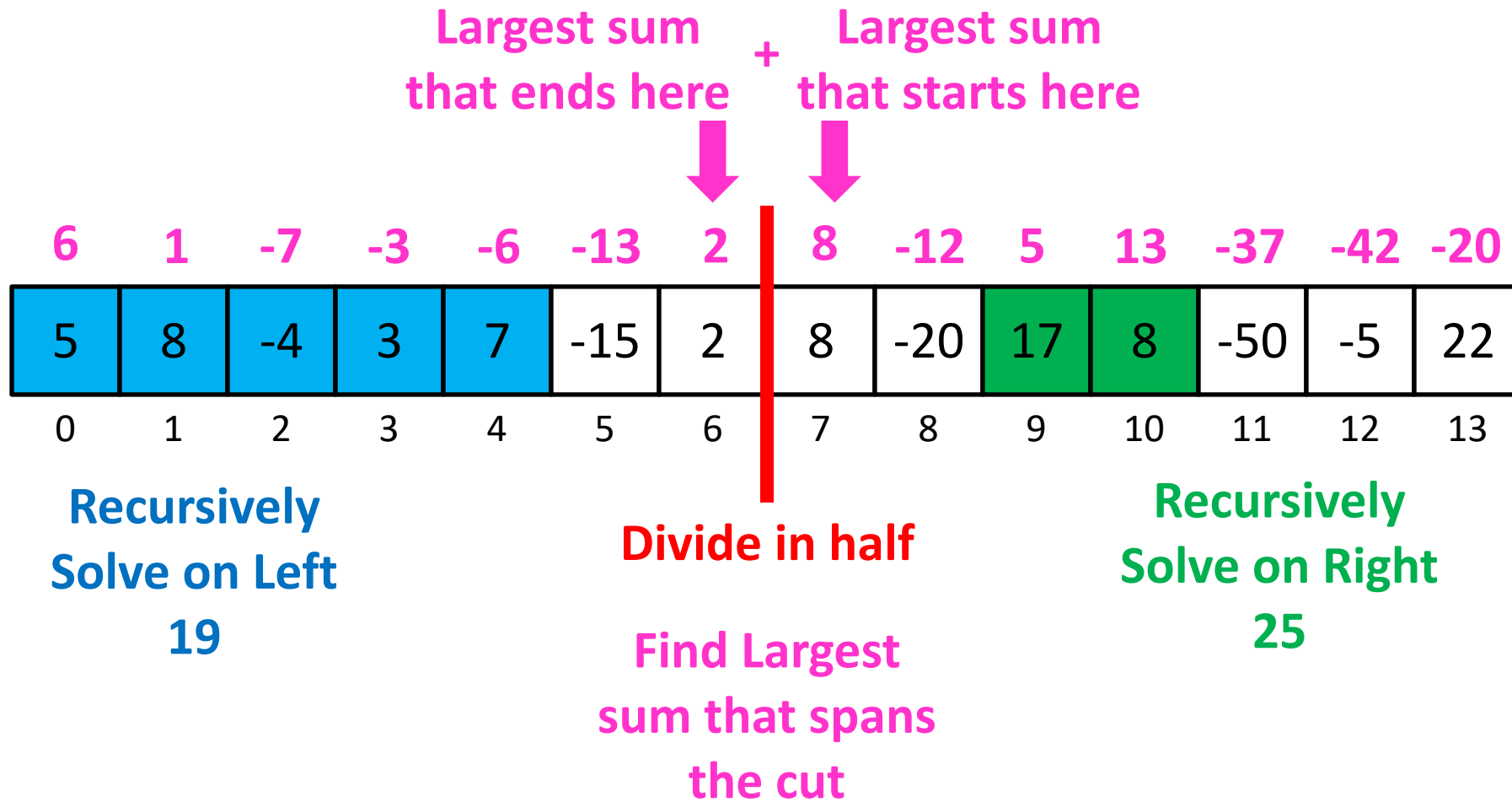
The maximum-sum subarray of a given array of integers A is the interval $[a, b]$ such that the sum of all values in the array between a and b inclusive is maximal.

Given an array of n integers (may include both positive and negative values), give a $O(n \log n)$ algorithm for finding the maximum-sum subarray.

Divide and Conquer $\Theta(n \log n)$

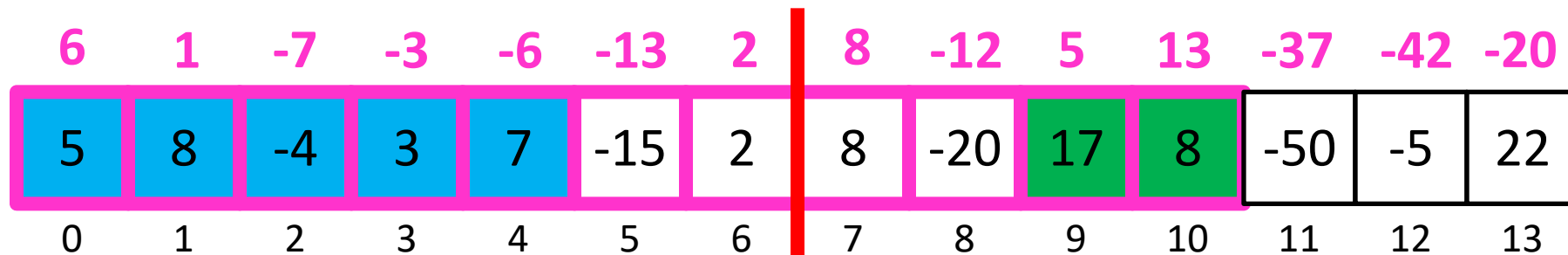


Divide and Conquer $\Theta(n \log n)$



Divide and Conquer $\Theta(n \log n)$

Return the Max of
Left, Right, Center



Recursively
Solve on Left
19

Divide in half

Find Largest
sum that spans
the cut
19

Recursively
Solve on Right
25

$$T(n) = 2T\left(\frac{n}{2}\right) + n$$

Divide and Conquer Summary

Typically multiple subproblems.
Typically all roughly the same size.

- **Divide**
 - Break the list in half
- **Conquer**
 - Find the best subarrays on the left and right
- **Combine**
 - Find the best subarray that “spans the divide”
 - I.e. the best subarray that ends at the divide concatenated with the best that starts at the divide

Generic Divide and Conquer Solution

```
def myDCalgo(problem):  
    if baseCase(problem):  
        solution = solve(problem) #brute force if necessary  
        return solution  
    subproblems = Divide(problem)  
    for sub in subproblems:  
        subsolutions.append(myDCalgo(sub))  
    solution = Combine(subsolutions)  
    return solution
```

MSCS Divide and Conquer $\Theta(n \log n)$

```
def MSCS(list):  
    if list.length < 2:  
        return list[0]    #list of size 1 the sum is maximal  
    {listL, listR} = Divide (list)  
    for list in {listL, listR}:  
        subSolutions.append(MSCS(list))  
    solution = max(solnL, solnR, span(listL, listR))  
    return solution
```

Types of “Divide and Conquer”

- **Divide and Conquer**
 - Break the problem up into several subproblems of roughly equal size, recursively solve
 - E.g. Karatsuba, Closest Pair of Points, Mergesort...
- **Decrease and Conquer**
 - Break the problem into a single smaller subproblem, recursively solve
 - E.g. Impossible Missions Force (Double Agents), Quickselect, Binary Search

Pattern So Far

- Typically looking to divide the problem by some fraction ($\frac{1}{2}$, $\frac{1}{4}$ the size)
- Not necessarily always the best!
 - Sometimes, we can write faster algorithms by finding **unbalanced** divides.