

# Non-Comparison Sorting Algorithms

CSE 332 Summer 2021

**Instructor:** Kristofer Wong

## Teaching Assistants:

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Rahul Misal      Winston Jodjana

# Announcements

- ❖ Midterm
- ❖ P1 Grades out. Regrade Requests
- ❖ Ex7: videos tonight!
  - Reflection q's cannot be edited after submissions came in, so a new reflection assignment will be posted. ***Disregard the reflections section in the Exercise 7 gradescope assignment. A new one will be posted!***
- ❖ Office Hours today pushed to 11:10 so I can release midterm
- ❖ Reminder: lecture activities deadline changed to 7am
- ❖ Amendment to Insertion/Selection Sort stability

# Lecture Outline

## ❖ QuickSort

- Comparison with MergeSort

## ❖ Comparison Sort Lower Bound

## ❖ Non-Comparison Sorts

- BucketSort
- RadixSort

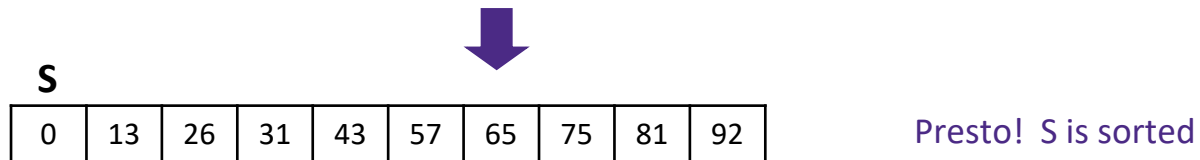
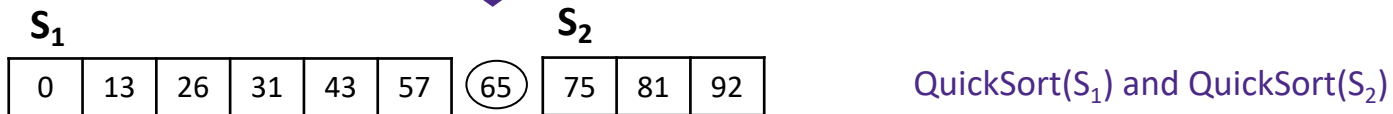
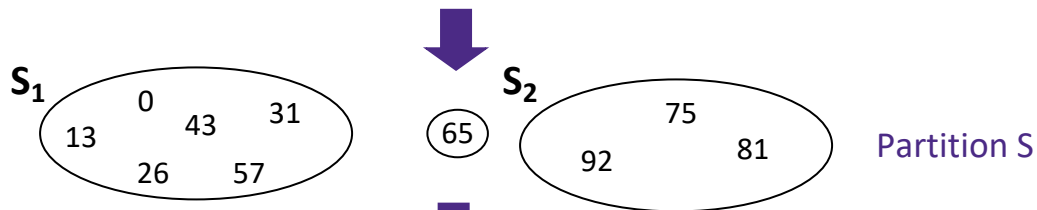
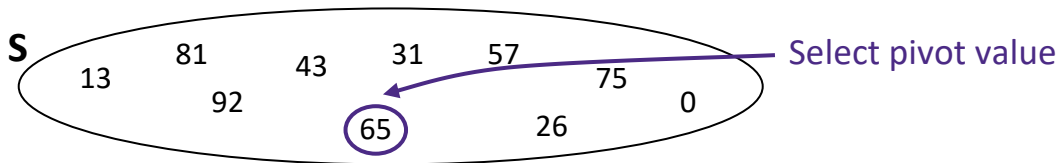
## ❖ Sorting Wrapup

# QuickSort Steps

1. Pick the pivot value(s)
  - Any choice is correct; data will end up sorted
  - For efficiency, these value(s) ought to approximate the median
  
2. Partition all the values into:
  - a. The values less than the pivot(s)
  - b. The pivot(s)
  - c. The values greater than the pivot(s)
  - d. .. In linear time? In-place? Stably?
  
3. Recursively QuickSort(A) and QuickSort(C)

✦✦ TA-DA! ✦✦

# QuickSort Intuition: Set Partitioning



# QuickSort Steps

1. Pick the pivot value(s)
  - Any choice is correct; data will end up sorted
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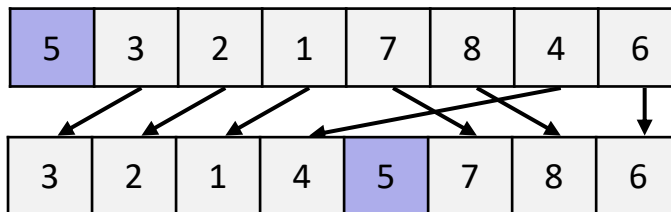
✨TA-DA!✨

## Recursive Call (1 of 3)

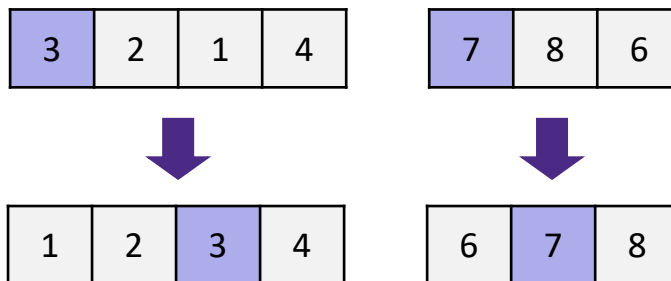
**Note:** for the remainder of this section, our pivot-selection algorithm is “first item in the subarray”

❖ After partitioning on 5:

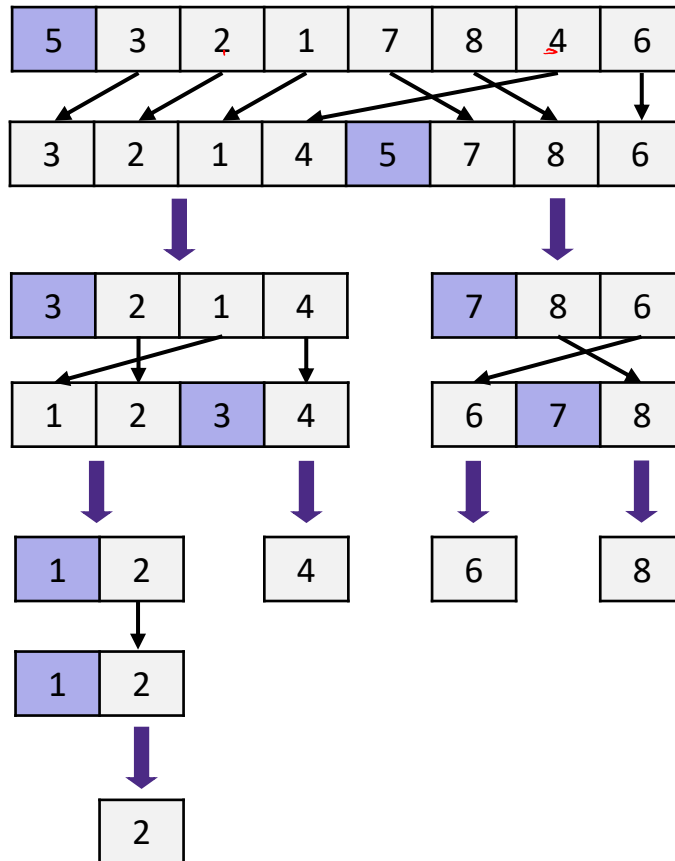
- 5 is in its “correct place” (ie, where it’d be if the array were sorted)



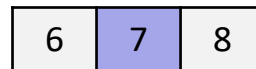
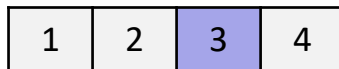
- Can now sort two halves separately (eg, through recursive use of partitioning)



## Recursive Call (2 of 3)



## Recursive Call (3 of 3)



# QuickSort Steps

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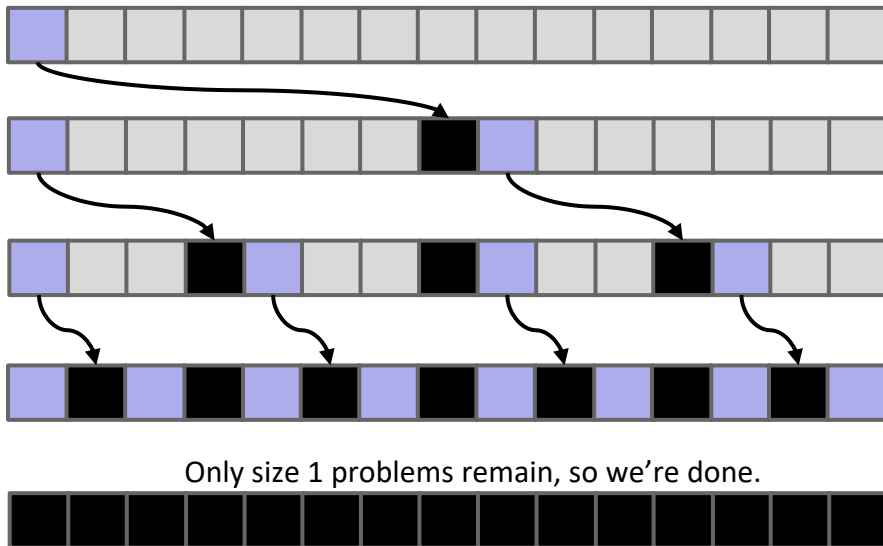
## 2. Partition all the values into:

- a. The values less than the pivot(s)
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## 3. Recursively QuickSort(A) and QuickSort(C)

✨TA-DA!✨

# Pivot Selection: Pivot is the Median



$$T(0) = T(1) = c_1$$

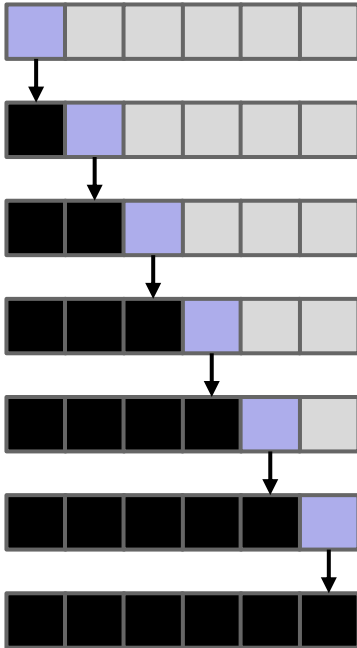
$$T(n) = 2T(n/2) + c_2 n$$

(partition is linear-time)

Same recurrence as MergeSort:

$$O(n \log n)$$

# Pivot Selection: Pivot is the Min/Max



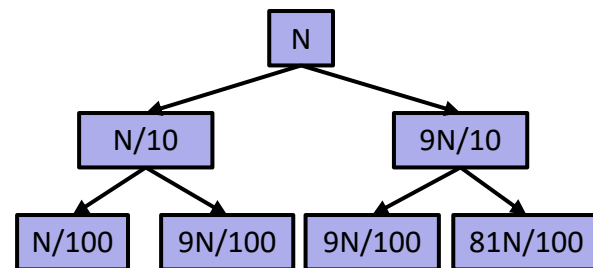
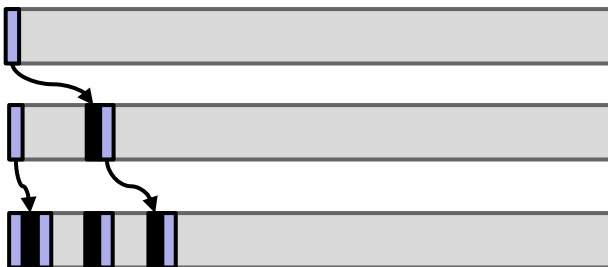
$$T(0) = T(1) = c_1$$

$$T(n) = T(n-1) + c_2n$$

Basically same recurrence as  
SelectionSort:  $O(n^2)$


# Pivot Selection: Pivot is Random

- ❖ Suppose pivot always ends up *at least 10% from either edge*



- ❖ Work at each level:  $O(N)$  and Runtime is  $O(NH)$ 
  - Height is approximately  $\log_{10/9} N = O(\log N)$
- ❖ Runtime:  $O(N \log N)$ 
  - See proof in text

# Pivot Selection Dictates Runtime!

- ❖ If pivot lands “somewhere good”, Quicksort is  $\Theta(N \log N)$
- ❖ However, the very rare  $\Theta(N^2)$  cases do happen in practice 
  - **Bad ordering:** Array already in (almost-)sorted order and pivot is first or last index
  - **Bad elements:** Array with all duplicates
- ❖ Three philosophies for avoiding worst-case behavior:
  1. **Randomness:** pick a random pivot; shuffle before sorting
    - Elegant, but (pseudo)random number generation can be slow
  2. **Smarter Pivot Selection:** calculate or approximate the median
    - Median-of-3: median of `arr[l0]`, `arr[hi-1]`, `arr[(hi+l0)/2]`
  3. **Introspection:** switch to safer sort if recursion goes too deep

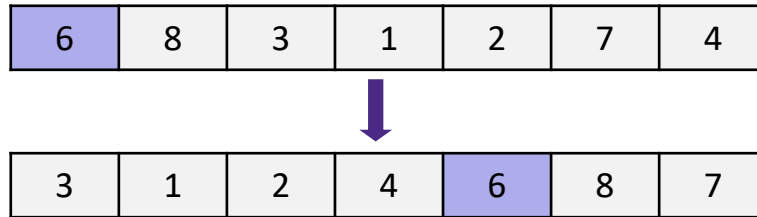
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# Partitioning: Problem Statement

- ❖ Given an array of elements and the 0<sup>th</sup> value as the pivot, write pseudocode that partitions the array



- ❖ Constraints:
  - Must complete in  $O(N \log N)$  time, but ideally  $\Theta(N)$
  - Must use  $O(N)$  space, but ideally  $\Theta(1)$
  - May use any data structure (eg, BSTs, stacks/queues, etc)
  - Ideally, preserves the elements' relative ordering ("stable")
- ❖ Conceptually simple, but hardest part to code up correctly!

# Partitioning: Option 1: Three-Pass

- ❖ Overview:
  - Copy “less than”s, then copy pivot(s), finally copy “greater-than”s
- ❖ Stable! 😊
- ❖ Uses extra space 😞
- ❖ Constants aren't great; very slow 😞

# Stable Three-Pass Partition

- Allocate an extra array to store output
- Loop: copy less-than-pivot values into output array
- Loop: copy pivot value(s) into output array
- Loop: copy greater-than-pivot values into output array

5	550	10	4	10	9	330
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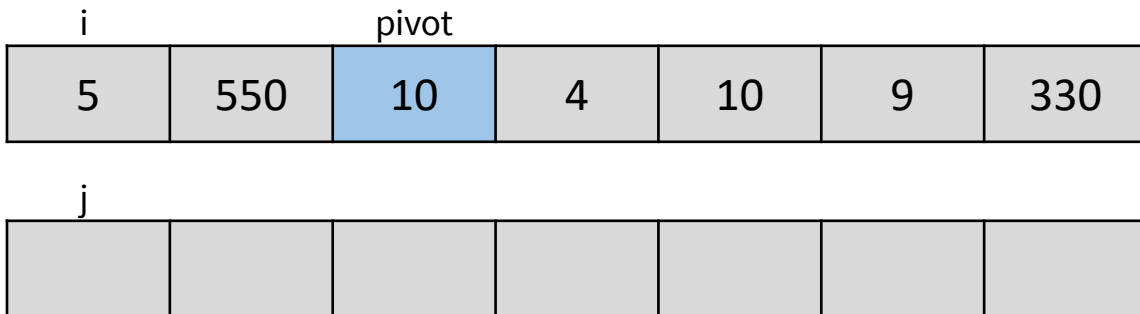
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*For demo purposes, we are using 10 as our pivot. However, there are many pivot-selection algorithms, including picking the 0th element.*

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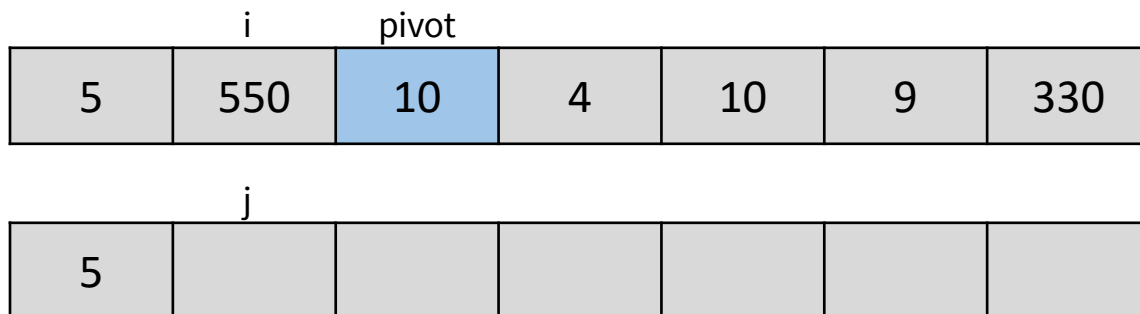
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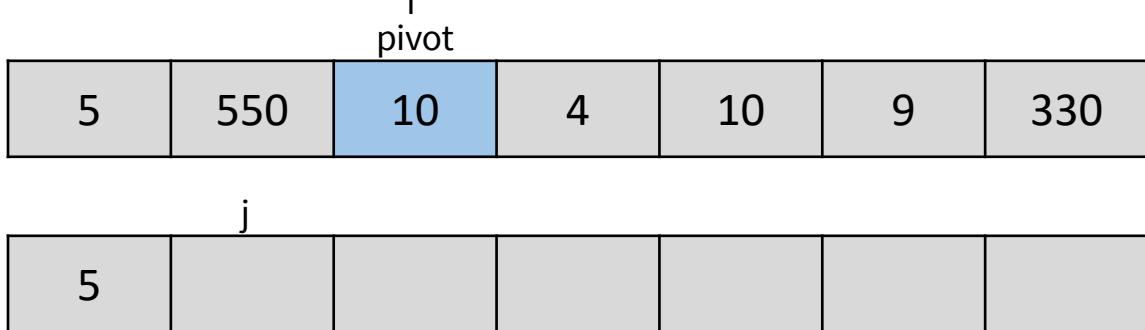
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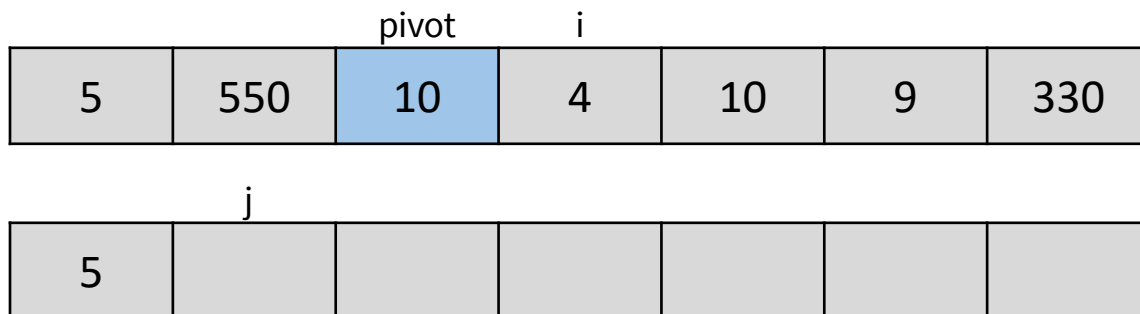
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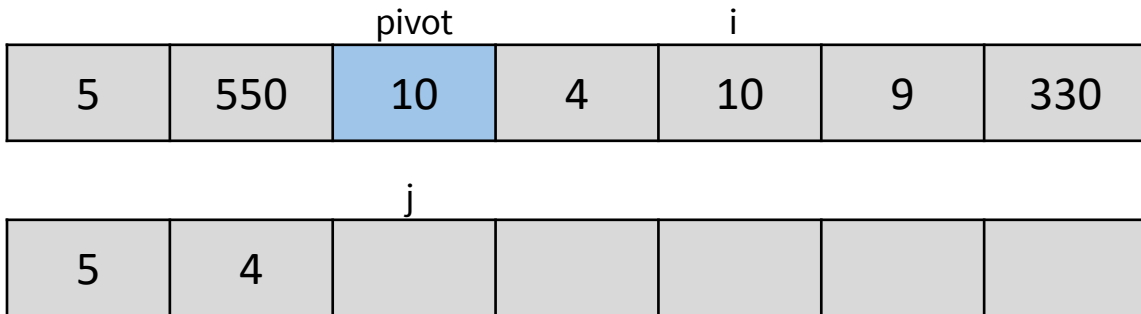
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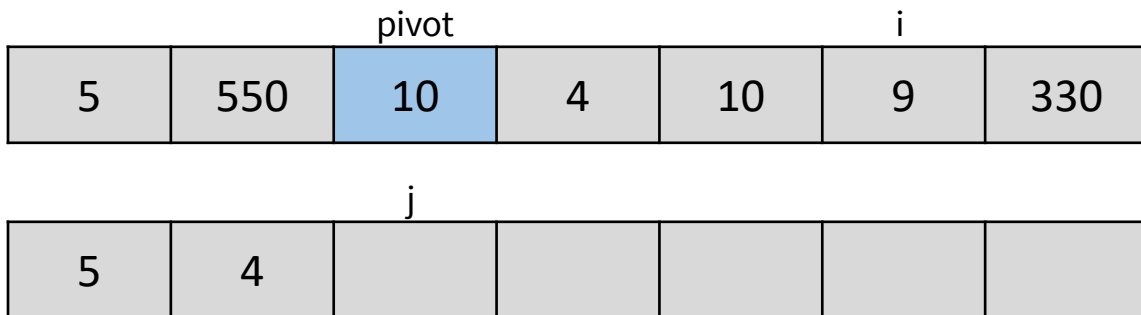
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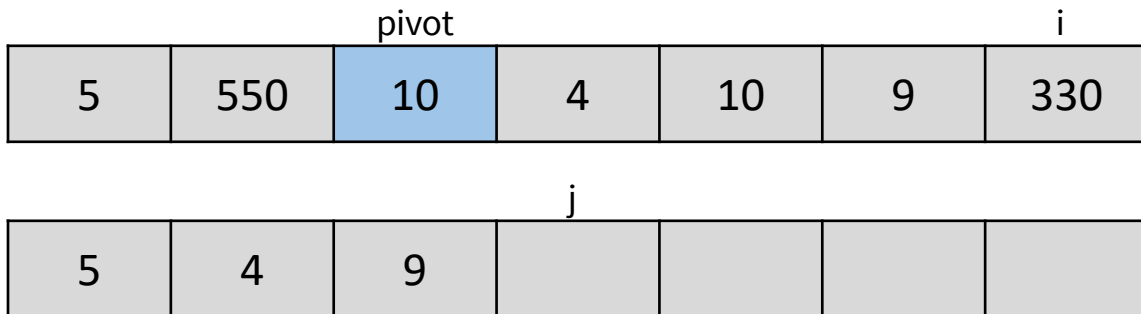
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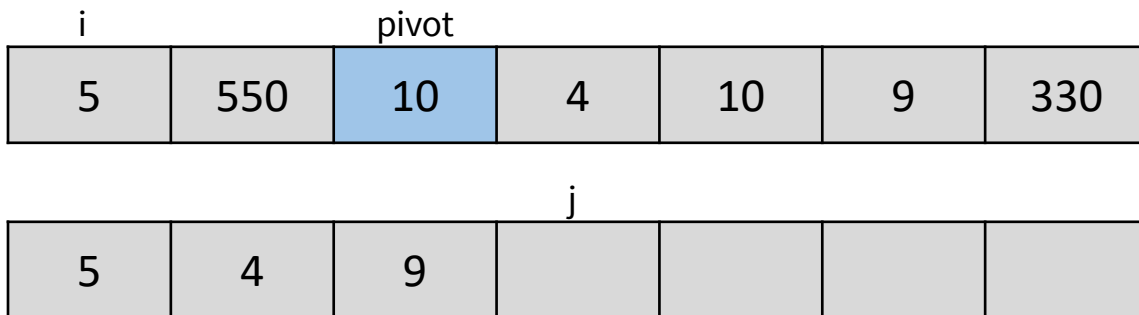
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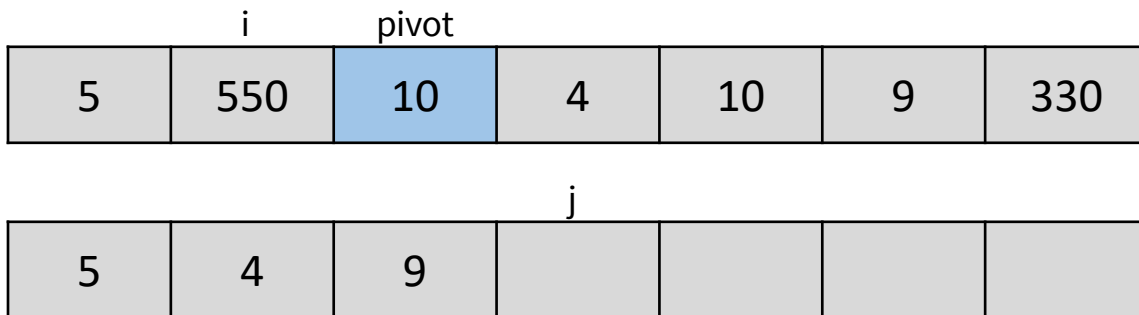
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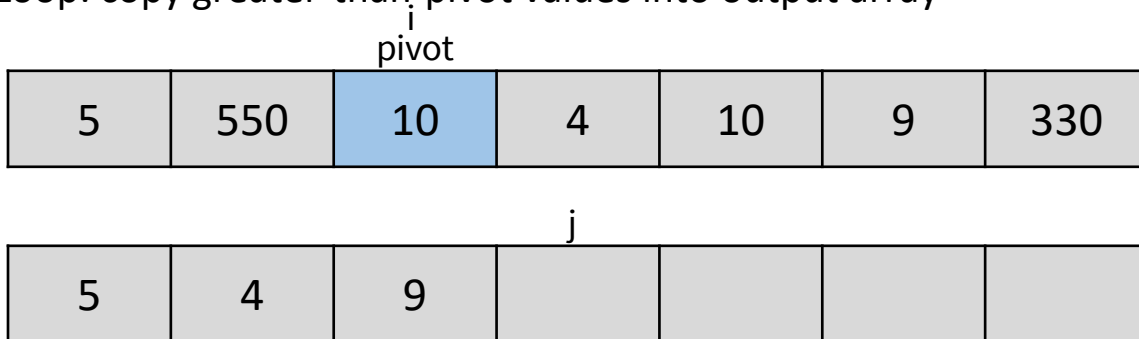
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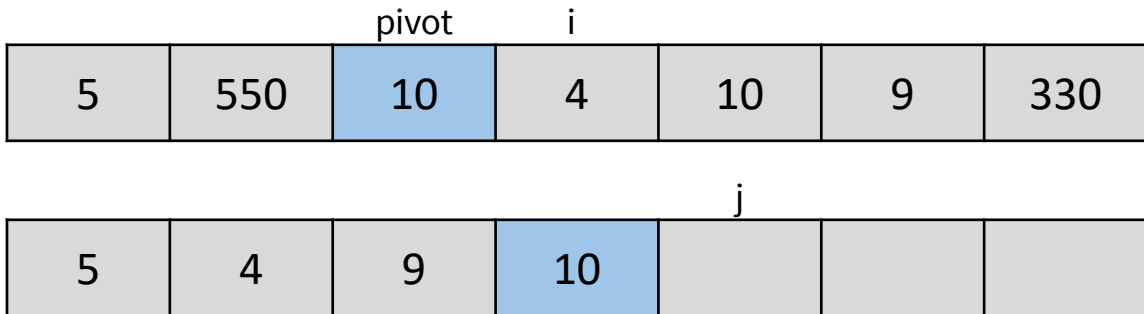
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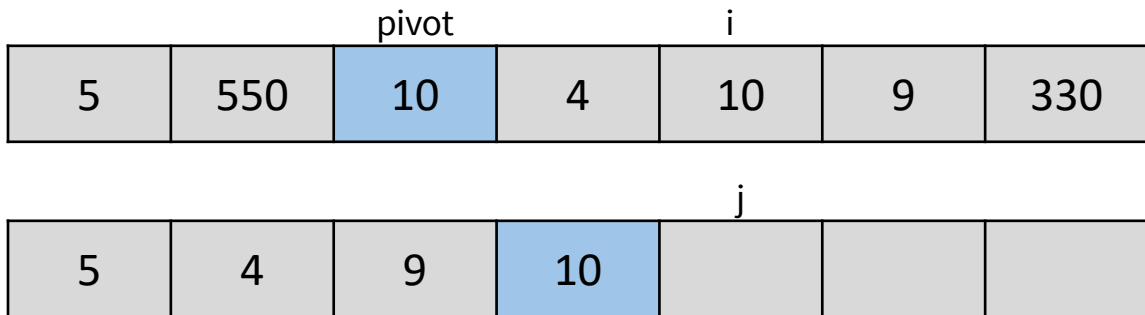
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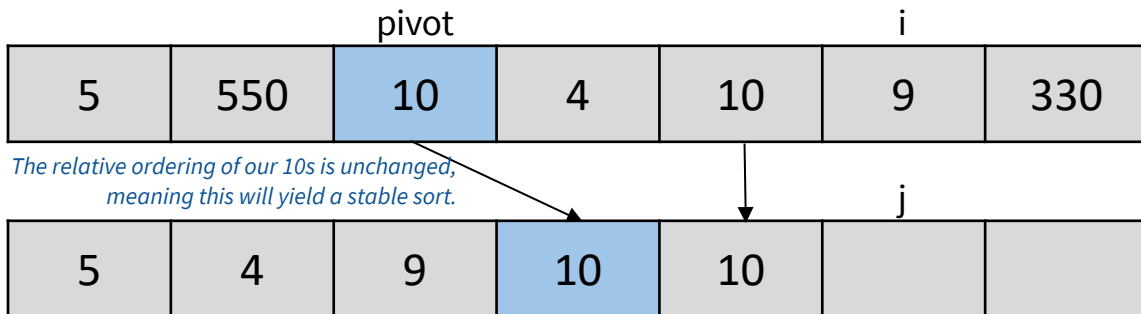
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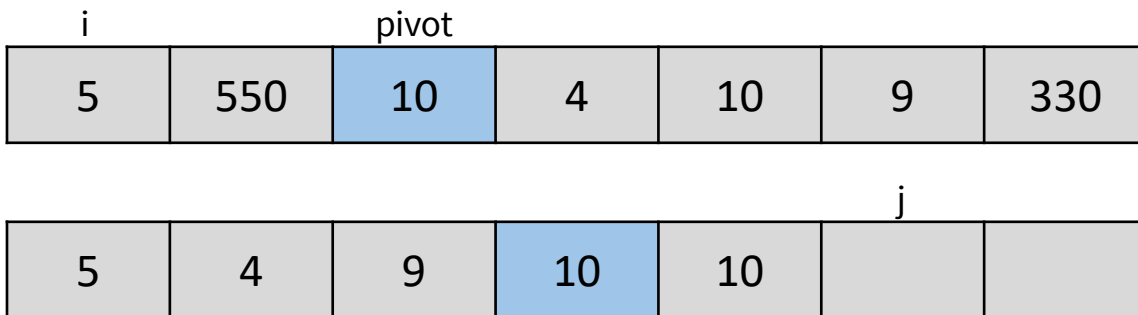
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		pivot				i
5	550	10	4	10	9	330

					j	
5	4	9	10	10		

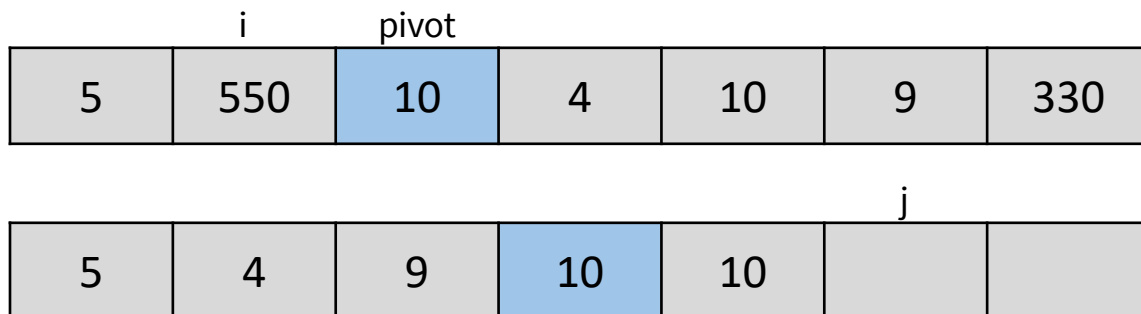
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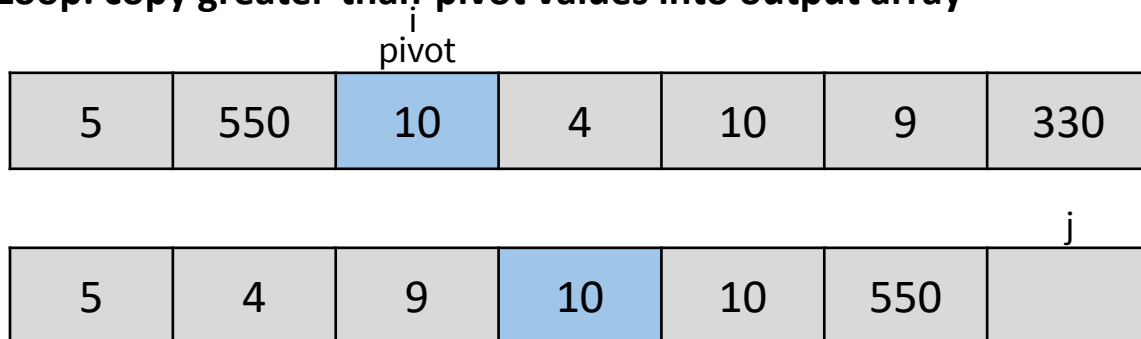
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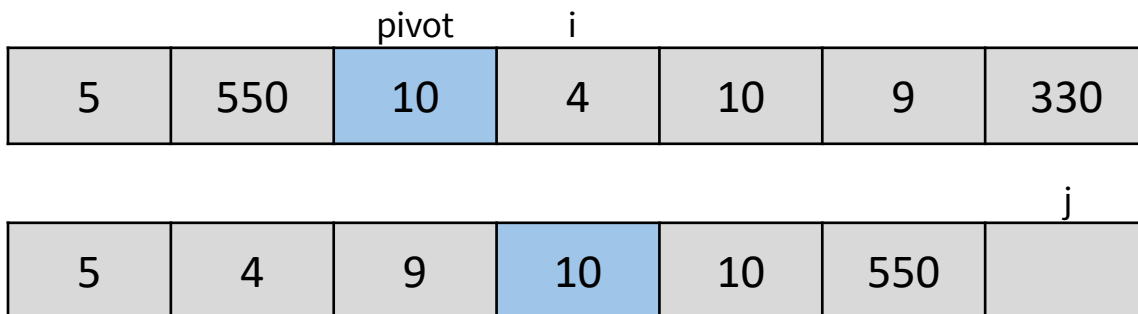
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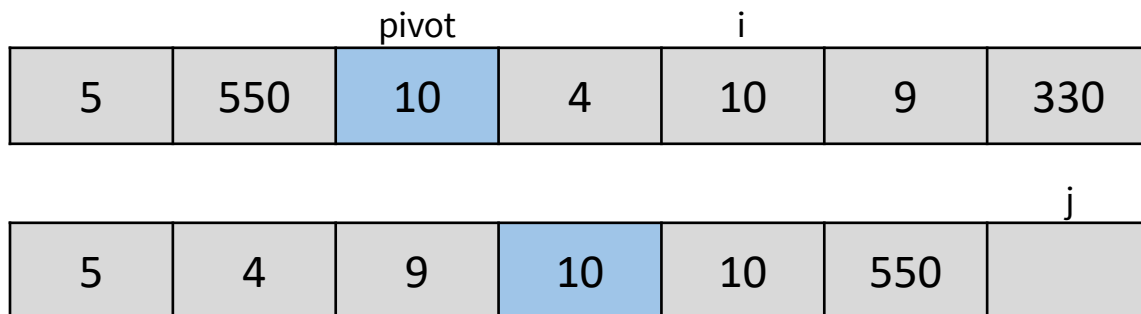
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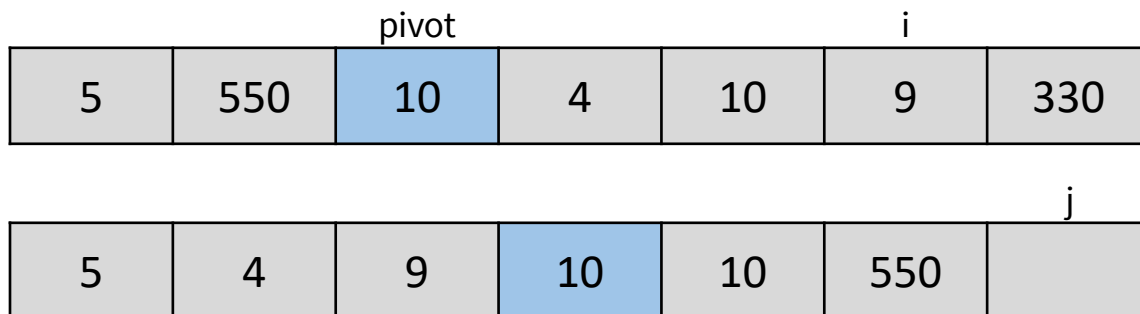
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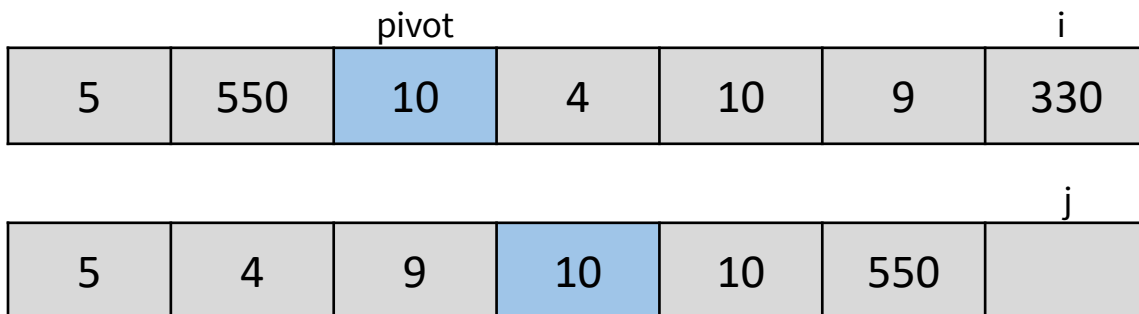
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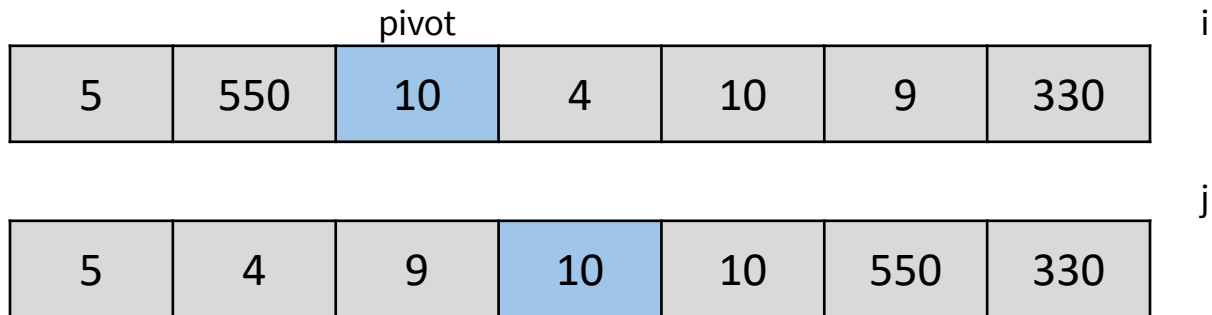
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*Done partitioning!*

# Partitioning: Option 2: Hoare Partitioning

- ❖ As published in Hoare's original QuickSort paper!
- ❖ Intuition:
  - L loves small items (i.e.,  $< \text{pivot}$ ) and R loves large items (i.e.,  $> \text{pivot}$ )
  - Walk towards each other, swapping anything they don't like
- ❖ Algorithm:
  1. Swap pivot with `arr[lo]` ("move it out of the way")
  2. Start `i` at `lo+1`, and `j` at `hi-1`
  3. Move `j` rightward until we hit value  $< \text{pivot}$  ("belongs on left")
  4. Move `i` leftward until we hit value  $> \text{pivot}$  ("belongs on right")
  5. Swap `arr[i]` and `arr[j]`
  6. When they meet, swap `arr[lo]` and `arr[i]` ("put pivot in correct place")

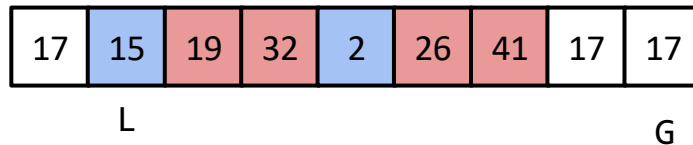
```
while (i < j)
    if (arr[j] > pivot) j--;
    else if (arr[i] <= pivot) i++;
    else swap(arr[i], arr[j])
```

# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
- G pointer is a friend to large items, and hates small or equal items.
- Walk pointers towards each other, stopping on a hated item.
  - When both pointers have stopped, swap and move pointers by one.
- When pointers cross, you are done.

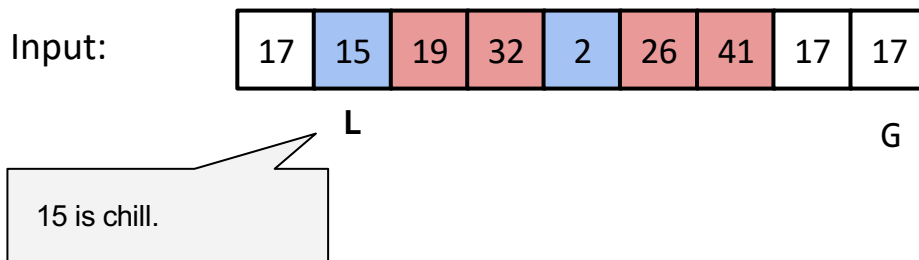
Input:



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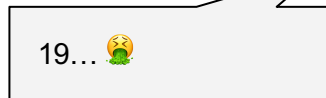
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Input:



L

G



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Input:



L

G

17 sucks

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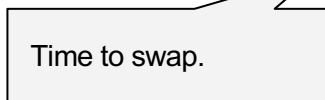
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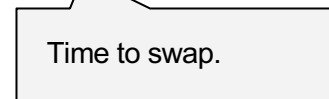
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L



G

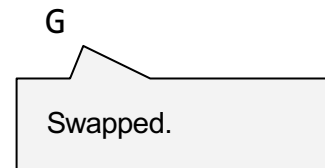
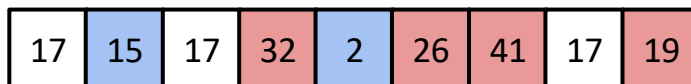


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- Walk pointers towards each other, stopping on a hated item.
  - **When both pointers have stopped, swap and move pointers by one.**
- When pointers cross, you are done.

Input:



# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
- G pointer is a friend to large items, and hates small or equal items.
- **Walk pointers towards each other, stopping on a hated item.**
  - When both pointers have stopped, swap and move pointers by one.
- When pointers cross, you are done.

Input:



L

G

32 is too big ☹

# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
- G pointer is a friend to large items, and hates small or equal items.
- **Walk pointers towards each other, stopping on a hated item.**
  - When both pointers have stopped, swap and move pointers by one.
- When pointers cross, you are done.

Input:



L

G

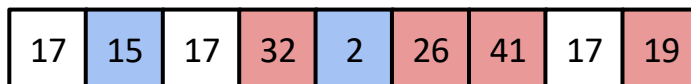
Too small

# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
- G pointer is a friend to large items, and hates small or equal items.
- Walk pointers towards each other, stopping on a hated item.
  - **When both pointers have stopped, swap and move pointers by one.**
  - When pointers cross, you are done walking.
- Swap pivot with G.

Input:



L

Time to swap.

G

Time to swap.

# Hoare Partitioning

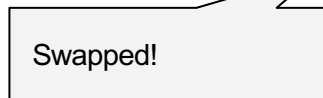
Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
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- Walk pointers towards each other, stopping on a hated item.
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  - When pointers cross, you are done walking.
- Swap pivot with G.

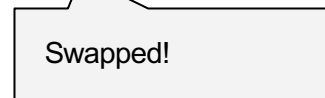
Input:



L



G



# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
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- **Walk pointers towards each other, stopping on a hated item.**
  - When both pointers have stopped, swap and move pointers by one.
  - When pointers cross, you are done walking.
- Swap pivot with G.

Input:



L

G

2 is cool.

# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
- G pointer is a friend to large items, and hates small or equal items.
- **Walk pointers towards each other, stopping on a hated item.**
  - When both pointers have stopped, swap and move pointers by one.
  - When pointers cross, you are done walking.
- Swap pivot with G.

Input:



L G

26 thicc. No thx

# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
- G pointer is a friend to large items, and hates small or equal items.
- **Walk pointers towards each other, stopping on a hated item.**
  - When both pointers have stopped, swap and move pointers by one.
  - When pointers cross, you are done walking.
- Swap pivot with G.

Input:



L

G

41 you good

# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
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- **Walk pointers towards each other, stopping on a hated item.**
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Input:



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- **Walk pointers towards each other, stopping on a hated item.**
  - When both pointers have stopped, swap and move pointers by one.
  - When pointers cross, you are done walking.
- Swap pivot with G.

Input:



G L

2 what u doin here 🤔  
also hi L!!

# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
- G pointer is a friend to large items, and hates small or equal items.
- Walk pointers towards each other, stopping on a hated item.
  - When both pointers have stopped, swap and move pointers by one.
  - **When pointers cross, you are done walking.**
- Swap pivot with G.

Input:



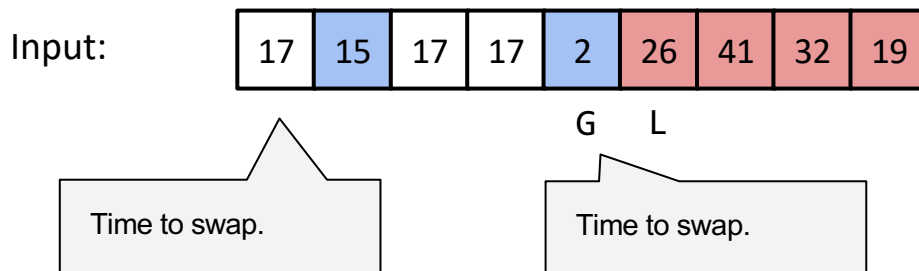
G L

2 what u doin here 🤔  
also hi L!!

# Hoare Partitioning

Create L and G pointers at left and right ends.

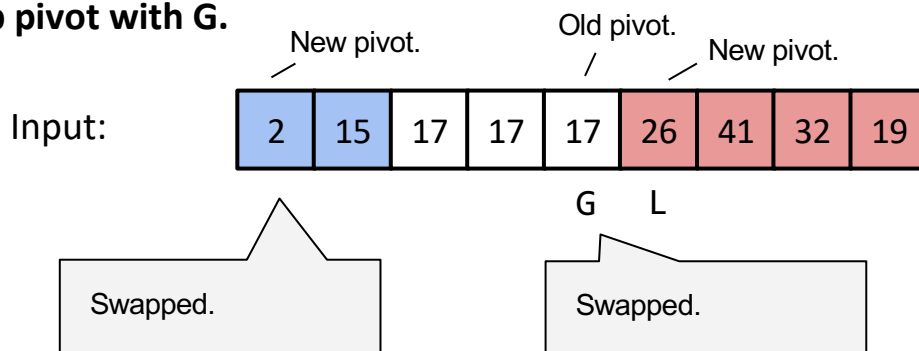
- L pointer is a friend to small items, and hates large or equal items.
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- Walk pointers towards each other, stopping on a hated item.
  - When both pointers have stopped, swap and move pointers by one.
  - When pointers cross, you are done walking.
- **Swap pivot with G.**



# Hoare Partitioning

Create L and G pointers at left and right ends.

- L pointer is a friend to small items, and hates large or equal items.
- G pointer is a friend to large items, and hates small or equal items.
- Walk pointers towards each other, stopping on a hated item.
  - When both pointers have stopped, swap and move pointers by one.
  - When pointers cross, you are done walking.
- **Swap pivot with G.**



- ❖ Partition the following array using Hoare's partitioning algorithm
  - The pivot, 5, has already been moved to the front
  - Sort only by the numbers (eg, 1); the extra letter (eg, 1a) is to help you determine stability

5	1a	6	9	3	7	2	4a	4b	1b
---	----	---	---	---	---	---	----	----	----

Swap pivot with `arr[lo]` ("move it out of the way")

Start `i` at `lo+1`, and `j` at `hi-1`

Move `j` rightward until we hit value `< pivot` ("belongs on left")

Move `i` leftward until we hit value `> pivot` ("belongs on right")

Swap `arr[i]` and `arr[j]`

When they meet, swap `arr[lo]` and `arr[i]` ("put pivot in correct place")

```
while (i < j)
  if (arr[j] > pivot) j--;
  else if (arr[i] <= pivot) i++;
  else swap(arr[i], arr[j])
```

# Notes on Hoare Partitioning

- Unstable 😞
- Good constants: single-pass and in-place 😊

# Partitioning: Option 3: Three-Way

- ❖ Pick *two* pivots
  - Same intuition as median-of-three: it's hard to pick multiple bad pivots simultaneously
- ❖ Like Hoare Partitioning, use two pointers walking to the middle
  - But split array into three pieces, not two
  - Good constants: single-pass and in-place;  $\log_3 N$  vs  $\log_2 N$  😊
  - Still an unstable sort 😞
- ❖ Used in Java's `Arrays.sort()`, Python's unstable sort, etc
  - Basically the de-facto partition algorithm circa 2020
- ❖ Won't be going through this version

# QuickSort Steps

1. Pick the pivot value(s)
  - Any choice is correct; data will end up sorted
  - For efficiency, these value(s) ought to approximate the median
  
2. Partition all the values into:
  - a. The values less than the pivot(s)
  - b. The pivot(s)
  - c. The values greater than the pivot(s)
  - d. ... In linear time? In-place? Stably?
  
3. Recursively QuickSort(A) and QuickSort(C)

✦✦ TA-DA! ✦✦

# QuickSort: End-to-end Example (1 of 3)

1. Pick pivot (we'll use median-of-3)

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

2. Partition (we'll use Hoare Partitioning)
  - Move pivot to the beginning position

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

- Let  $lo = 1$  and  $hi = 9$ ; loop until we find "swappable" values

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



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

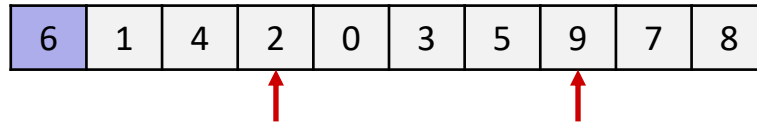


6	1	4	9	0	3	5	2	7	8
---	---	---	---	---	---	---	---	---	---

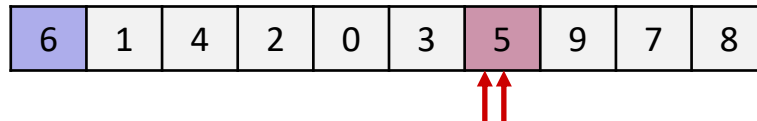
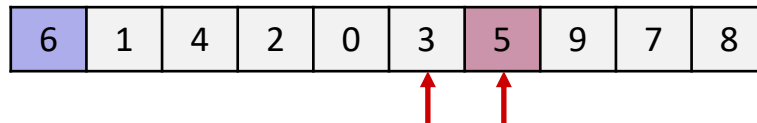
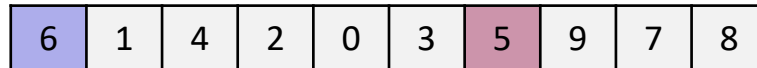


## QuickSort: End-to-end Example (2 of 3)

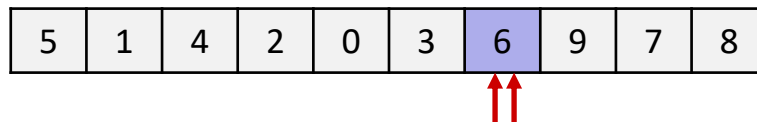
- Swap  $l_o = 3$  and  $h_i = 7$



- Keep looping

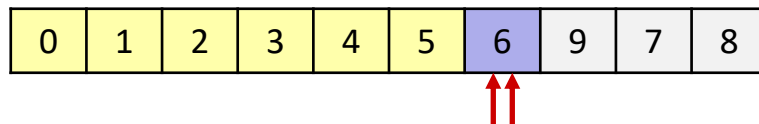


- Done! Swap pivot into position

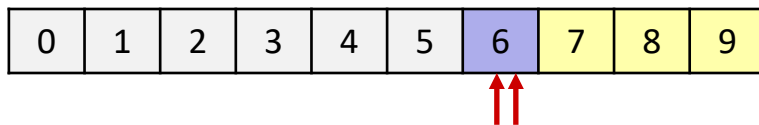


## QuickSort: End-to-end Example (3 of 3)

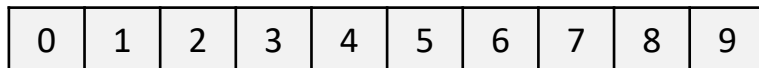
3. Recursively sort left (0 to  $lo-1 = 5$ )



4. Recursively sort right ( $hi+1 = 7$  to `arr.length`)

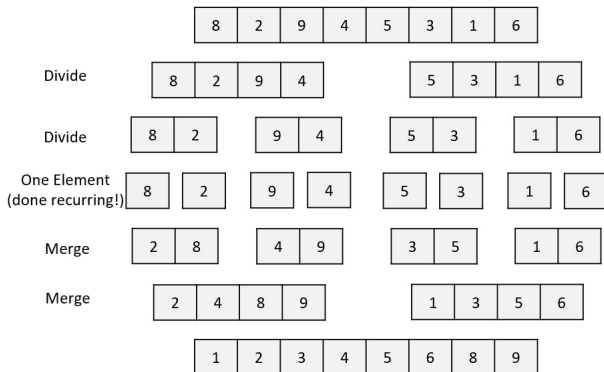


5. Sorted!



# QuickSort Optimization: Cutoffs (1 of 2)

- ❖ For small  $n$ , recursion tends to cost more than a quadratic sort
  - Remember: asymptotic complexity applies to large  $n$
  - Recursive calls add overhead (which “isn’t worth it” for small  $n$ )
  
- ❖ Recursive calls for small  $n$  are the most common (“leaf calls”)
  - Calls for small  $n$  are the vast majority of the recursive calls!
  - Image below is for mergesort, but shows this well



## QuickSort Optimization: Cutoffs (2 of 2)

- ❖ So, switch algorithms for subproblems below a **cutoff** size
  - Eg, Java 12 uses InsertionSort for primitive types when  $n < 47$

```
void quickSort(int[] arr, int lo, int hi) {  
    if (hi - lo < CUTOFF)  
        insertionSort(arr, lo, hi);  
    else  
        ...  
}
```

- ❖ Switching algorithms after a cutoff is a common technique!
  - E.g. *parallel* algorithms switch to *sequential* after a certain cutoff
  - E.g. MergeSort also uses cutoffs to switch to InsertionSort
- ❖ Does not affect asymptotic complexity, just the constants

# QuickSort vs MergeSort (1 of 2)

*Different algorithms, same problem*

## ❖ MergeSort:

### ■ Execution

- Does its work “on the way up”
  - i.e., in the merge, after the recursive call returns
- Uses its auxiliary space very effectively:
  - Works well on linked lists
  - Linear merges minimize disk accesses

### ■ Time: always $O(n \log n)$

## ❖ QuickSort:

### ■ Execution:

- Does its work “on the way down”
  - i.e., in the partition, before the recursive call
- Doesn’t need auxiliary space

### ■ Runtime: $O(n \log n)$ in best and randomized cases 😊

- But  $O(n^2)$  worst-case ☹

# QuickSort vs MergeSort (2 of 2)

- ❖ Asymptotic Runtime:
  - QuickSort is  $O(n \log n)$  in best and randomized cases, but  $O(n^2)$  worst-case
  - MergeSort is always  $O(n \log n)$
- ❖ Constants Matter!
  - QuickSort does fewer copies and more comparisons, so it depends on the relative cost of these two operations
  - Typically, cost of copies is higher so QuickSort really *is* the “quickest”

# Lecture Outline

- ❖ QuickSort
  - Comparison with MergeSort
- ❖ **Comparison Sort Lower Bound**
- ❖ Non-Comparison Sorts
  - BucketSort
  - RadixSort
- ❖ Sorting Wrapup

# A Different View of Sorting

❖ Assume we have  $n$  elements, none are equal (ie, no duplicates)

- - **$n!$  permutations** (possible orderings) of the elements. For  $n=3$

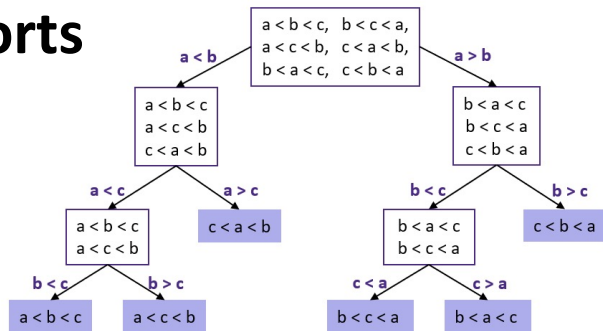
$a < b < c$	$a < c < b$	$b < a < c$	$b < c < a$	$c < a < b$	$c < b < a$
-------------	-------------	-------------	-------------	-------------	-------------

❖ Assume an “OptimalSort”

- Instead of describing how it **works**, we’ll describe what it **knows and when it knows it**
- Starts “knowing nothing”; “anything is possible”
- Each binary:  $a < b$  or  $b < a$  comparison gains information, eliminating some possibilities
  - Each comparison eliminates (at most) half of remaining possibilities
- In the end, narrows down to a single possibility

# Representing Comparison Sorts

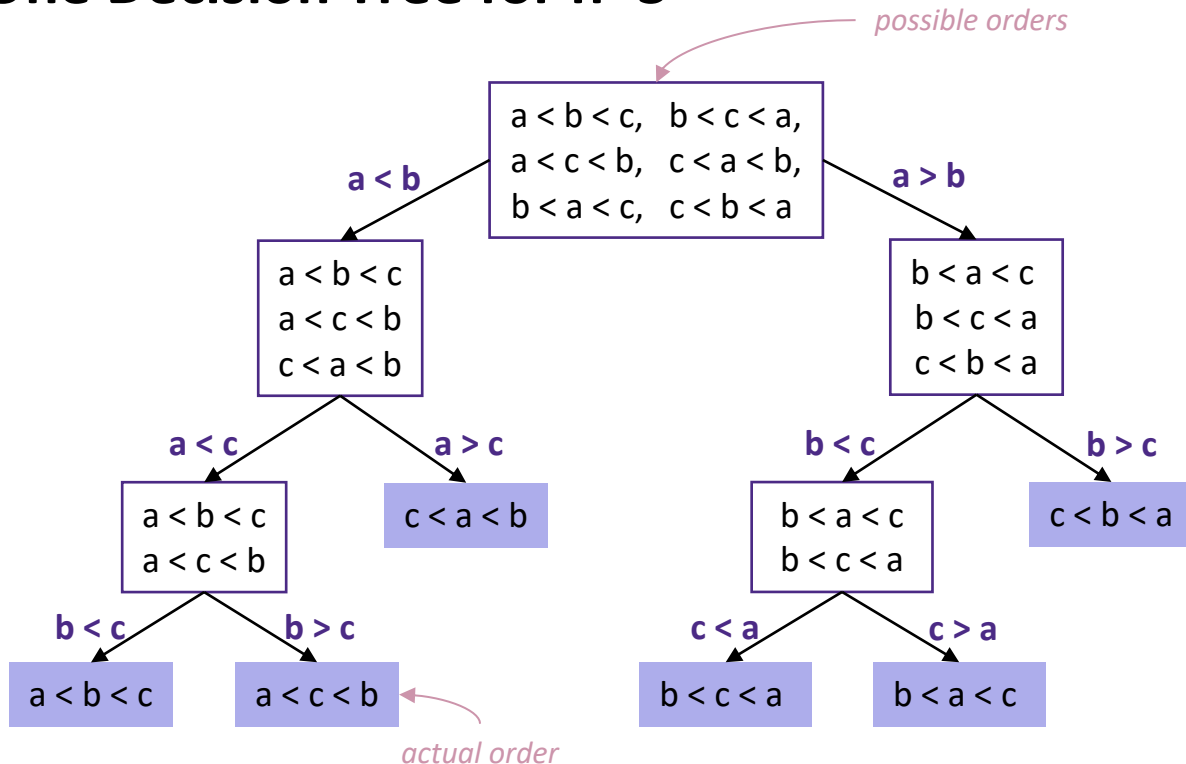
- ❖ Let's represent these binary comparisons as a binary tree!



- ❖ Called a *Decision Tree*
  - Nodes contain “set of remaining possible orderings”
  - The root contains all possible orderings; anything is possible
  - The leaves contain exactly one specific ordering
  - Edges are “answers from a comparison”

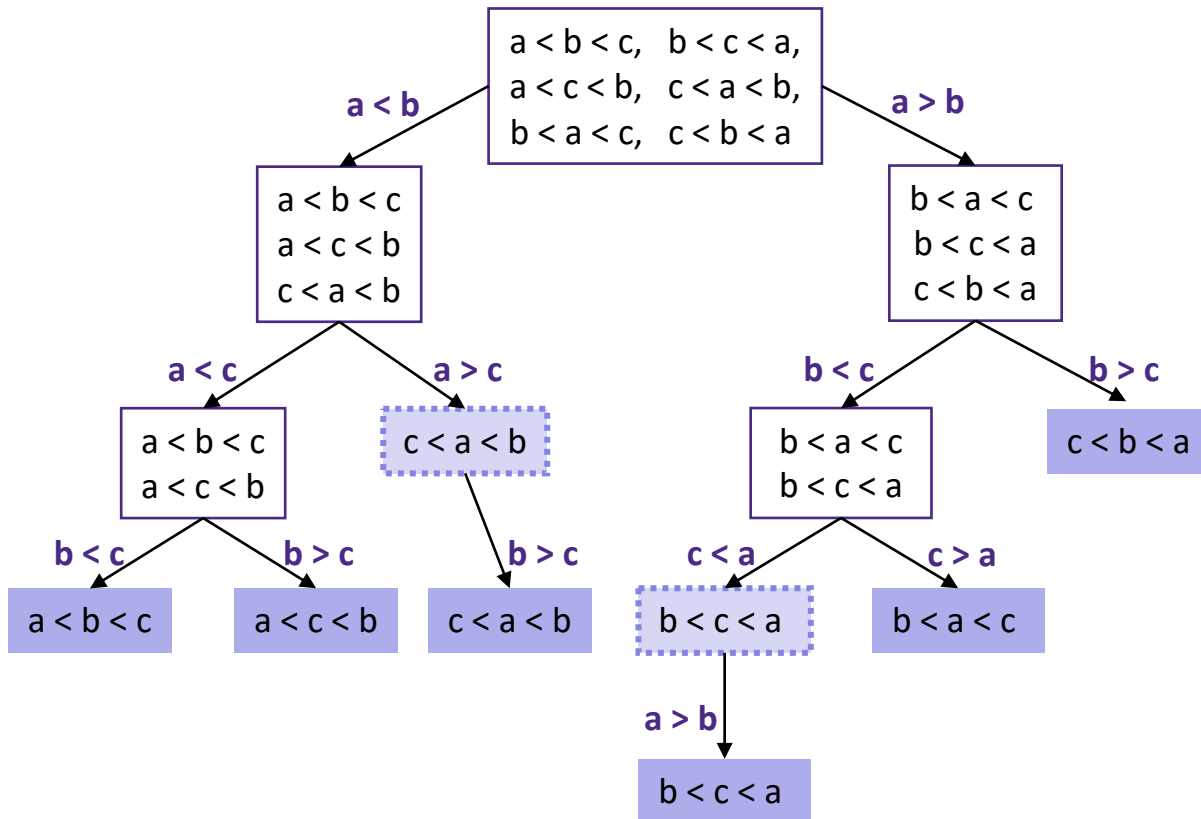
*We are not actually building the tree; it's what our proof uses to represent “the most the algorithm could know so far”*

# One Decision Tree for $n=3$



- The leaves contain all the possible orderings of  $a, b, c$
- A different algorithm would lead to a different tree

# Another Decision Tree for n=3



# What the Decision Tree Tells Us

- ❖ Because any order is possible, any algorithm needs to ask enough questions to produce all  $n!$  leaves (ie, orderings)
  - Each answer/ordering may lead to a different leaf
  - ***So the binary tree must be big enough to have  $n!$  leaves***
- ❖ Running *any* algorithm on *any* input will at best correspond to a root-to-leaf path in *some* decision tree with  $n!$  leaves
  - Path length is the number of comparison operations needed
  - ***So no algorithm can have worst-case running time better than the height of a tree with  $n!$  leaves***
    - Because the worst-case number-of-comparisons for an algorithm is an input that yields to a longest path in algorithm's decision tree

How did we get  $n!$ ?

This is a 312 question. We have  $n$  variables and we wish to know how many different ways they can be ordered. In other words, want the number of permutations of size  $n$ . A permutation is the choice of  $r$  things from a set of  $n$  things without replacement and where the order matters:

$${}^n P_r = \frac{n!}{(n-r)!}$$

In our case,  $r = n$  because we want the choice of all  $n$  things from the set of  $n$  things, so:  ${}^n P_n = \frac{n!}{(n-n)!} = \frac{n!}{0!} = \frac{n!}{1} = n!$

## Lower Bound on Height (1 of 2)

- ❖ A binary tree of height  $h$  has at most how many leaves?

$$L \leq \underline{\hspace{2cm}}$$

- ❖ A binary tree with  $L$  leaves has height at least:

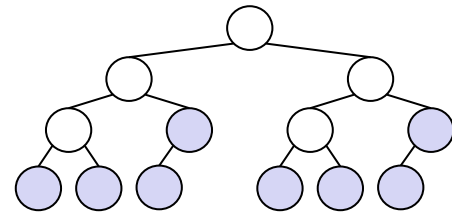
$$h \geq \underline{\hspace{2cm}}$$

- ❖ The decision tree has how many leaves:  $\underline{\hspace{2cm}}$

- ❖ So the decision tree has height:

$$h \geq \underline{\hspace{2cm}}$$

## Lower Bound on Height (2 of 2)



❖ The height of a binary tree with  $L$  leaves is at least  $\log_2 L$

❖ So the height of our decision tree,  $h$ :

$$h \geq \log_2 (n!)$$

$$= \log_2 (n \cdot (n-1) \cdot (n-2) \cdots (2) \cdot (1))$$

$$= \log_2 n + \log_2 (n-1) + \dots + \log_2 1$$

$$\geq \log_2 n + \log_2 (n-1) + \dots + \log_2 (n/2)$$

$$\geq (n/2) \log_2 (n/2)$$

$$= (n/2)(\log_2 n - \log_2 2)$$

$$= (1/2)n \log_2 n - (1/2)n$$

$$\in \Omega (n \log n)$$

property of binary trees

definition of factorial

property of logarithms

keep first  $n/2$  terms

each of the  $n/2$  terms is  $\geq \log_2 (n/2)$

property of logarithms

arithmetic

So... No comparison based sort can be faster than  $n \log n$ .

# Lecture Outline

- ❖ QuickSort
  - Comparison with MergeSort
- ❖ Comparison Sort Lower Bound
- ❖ **Non-Comparison Sorts**
  - **BucketSort**
  - RadixSort
- ❖ Sorting Wrapup

# BucketSort (a.k.a. BinSort)

- ❖ If all values to be sorted are known to be integers between 1 and  $K$  (or any small range),
  - Create an array of size  $K$ , put each element in its **bucket (a.k.a. bin)**
  - If data is only integers, can store *count* of how many times that bucket has been used
- ❖ Output result via linear pass through array of buckets

count array	
1	
2	
3	
4	
5	

- Example:

$K=5$

Input: (5,1,3,4,3,2,1,1,5,4,5)

Output: 1,1,1,2,3,3,4,4,5,5,5

# Analyzing BucketSort

- ❖ Overall:  $O(n+K)$ 
  - Linear in  $n$ , but also linear in  $K$
  - $\Omega(n \log n)$  doesn't apply because **this is not a comparison sort**
- ❖ Good when range,  $K$ , is smaller (or not much larger) than  $n$ 
  - We don't spend time doing lots of comparisons of duplicates!
- ❖ Bad when  $K$  is much larger than  $n$ 
  - Wasted space; wasted time during final linear  $O(K)$  pass

# BucketSort with Data

- ❖ Most real lists aren't just #'s; we have data too
  - Make each bucket is a list (say, linked list)
  - To add to a bucket, place at end  $O(1)$  (keep pointer to last element)
    - Example: movie ratings (1=bad, ... 5=excellent)
    - Input=
      - 5: Pulp Fiction
      - 3: Harry Potter movies
      - 1: Star Wars I
      - 5: Star Wars V
    - Output= Star Wars I, Harry Potter movies, Pulp Fiction, Star Wars V

count array	
1	→ Star Wars V
2	
3	→ Harry Potter
4	
5	→ Pulp Fiction → Star Wars V

# Lecture Outline

- ❖ QuickSort
  - Comparison with MergeSort
- ❖ Comparison Sort Lower Bound
- ❖ Non-Comparison Sorts
  - BucketSort
  - **RadixSort**
- ❖ Sorting Wrapup

# RadixSort

- ❖ Radix = “the base of a number system”
  - Examples will use 10 because we are used to that
  - Implementations may use larger numbers
    - For example, for ASCII strings, might use 128
  
- ❖ Idea:
  - Bucket sort on one digit at a time
    - Number of buckets = radix
    - Starting with *least* significant digit, sort with Bucket Sort
    - Keeping sort *stable*
  - Do one pass per digit
  
- ❖ **Invariant:** After  $k$  passes, the last  $k$  digits are sorted
  
- ❖ **Aside:** Origins go back to the 1890 U.S. census

# RadixSort: Example (1 of 6)

0	1	2	3	4	5	6	7	8	9

**Input:** 333  
143  
591  
65  
332  
491

## First pass:

1. BucketSort by ones digit
2. Iterate thru and collect into a list
  - List is sorted by first digit

## RadixSort: Example (2 of 6)

0	1	2	3	4	5	6	7	8	9
	591 491	332	333 143		65				

**Input:** 333

143

591

65

332

491

**First pass:**

1. BucketSort by ones digit
2. Iterate thru and collect into a list
  - List is sorted by first digit

**Order is now:**

591

491

332

333

143

65

0	1	2	3	4	5	6	7	8	9
	591 491	332	333 143		65				

**Input:** 333  
143  
591  
65  
332  
491

**Second pass:**

1. BucketSort by tens digit, stably

**Order is now:**



0	1	2	3	4	5	6	7	8	9
	591 491	332	333 143		65				
0	1	2	3	4	5	6	7	8	9
			332 333	143		65			591 491

**Input:** 333  
143  
591  
65  
332  
491

**Second pass:**

1. BucketSort by tens digit, stably

Notice: if we chop off the 100's place, these are now sorted

**Order is now:**

332  
333  
143  
65  
591  
491

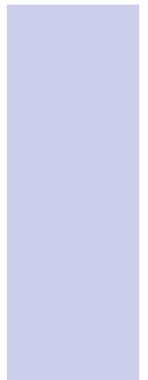
0	1	2	3	4	5	6	7	8	9
			332 333	143		65			591 491
0	1	2	3	4	5	6	7	8	9

**Input:** 333  
143  
591  
65  
332  
491

**Third pass:**

1. BucketSort by hundreds digit, stably

**Order is now:**



0	1	2	3	4	5	6	7	8	9
			332 333	143		65			591 491
0	1	2	3	4	5	6	7	8	9
65	143		332 333						491 591

**Input:** 333  
143  
591  
65  
332  
491

**Third pass:**

1. BucketSort by hundreds digit, stably

 *Only 3 digits; we're done!* 

**Order is now:**

65  
143  
332  
333  
491  
591

# Analysis of Radix Sort

- ❖ Performance depends on:
  - Input size:  $n$
  - Number of buckets = Radix:  $K$ 
    - e.g. Base 10 #: 10; binary #: 2; Alpha-numeric char: 62
  - Number of passes = “Digits”:  $P$ 
    - e.g. Ages of people: 3; Phone #: 10; Person’s name: ?
  
- ❖ Work per pass is 1 BucketSort: \_\_\_\_\_
  - *Each pass is a BucketSort!*
  
- ❖ Total work is \_\_\_\_\_
  - *We do ‘P’ passes, each of which is a BucketSort!*

# Comparison to Comparison Sorts

- ❖ Compared to comparison sorts, radix sorts are sometimes a win, but often not
  
- ❖ Example: Strings of English letters up to length 15
  - Approximate run-time:  $15 \cdot (52 + n)$
  - This is less than  $n \log n$  only if  $n > 33,000$
  - Of course, cross-over point depends on constant factors of the implementations plus  $P$  and  $B$ 
    - And radix sort can have poor locality properties
  
- ❖ Not really practical for many classes of keys
  - Strings: Lots of buckets

# Lecture Outline

- ❖ QuickSort
  - Comparison with MergeSort
- ❖ Comparison Sort Lower Bound
- ❖ Non-Comparison Sorts
  - BucketSort
  - RadixSort
- ❖ **Sorting Wrapup**

# Features of Sorting Algorithms

## ❖ In-place

- Sorted items occupy the same space as the original items. (No copying required, only  $O(1)$  extra space if any.)

## ❖ Stable

- Items in input with the same value end up in the same order as when they began.

# Sorting: Summary (1 of 3)

- ❖ Simple  $O(n^2)$  sorts can be fastest for small  $n$ 
  - SelectionSort, InsertionSort:
    - The latter is linear for mostly-sorted!
    - Good for “below a cut-off” to help divide-and-conquer sorts
  
- ❖ “Fancy”  $O(n \log n)$  sorts
  - HeapSort: not parallelizable
  - MergeSort: works as external sort
  - QuickSort:  $O(n^2)$  in worst-case; cost of comparisons/copies often makes it fastest

## Sorting : Summary (2 of 3)

- ❖  $\Omega(n \log n)$  is worst-case and average lower-bound for sorting by comparisons
- ❖ Non-comparison sorts
  - Bucket sort good for small number of key values
  - Radix sort uses fewer buckets and more phases
- ❖ Best way to sort? It depends!

# Sorting: Summary (3 of 3)

	Best-Case	Worst-Case	Randomized Case	In-Place?	Stable?	Notes
InsertionSort	$\Theta(N)$	$\Theta(N^2)$	$\Theta(N^2)$	Yes	Yes	Fastest for small or partially-sorted input
SelectionSort	$\Theta(N^2)$	$\Theta(N^2)$	$\Theta(N^2)$	Yes	No	
In-Place HeapSort	$\Theta(N)$	$\Theta(N \log N)$	$\Theta(N \log N)$	Yes	No	Slow in practice
MergeSort	$\Theta(N \log N)$	$\Theta(N \log N)$	$\Theta(N \log N)$	No	Yes	Fastest stable sort
QuickSort <i>(1st-element pivot + 3-pass partition)</i>	$\Theta(N \log N)$	$\Theta(N^2)$	$\Theta(N \log N)$	No	Yes	$\geq 2x$ slower than MergeSort
QuickSort <i>(Median-of-three pivot + Hoare partition + cutoffs)</i>	$\Omega(N)$	$O(N^2)$	$\Theta(N \log N)$	Yes	No	Fastest comparison sort
BucketSort	$\Theta(N+K)$	$\Theta(N+K)$	$\Theta(N+K)$	No	Yes	
RadixSort	$\Theta(P(K+N))$	$\Theta(P(K+N))$	$\Theta(P(K+N))$	No	Yes	