CSE 332: Sorting

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Announcements (2/3/14)

- Reading for this lecture: Chapter 7.
- HW 4 due Wednesday
  - no new HW out this week
- Midterm next Monday

Sorting

- Input
  - an array A of data records
  - a key value in each data record
  - a comparison function which imposes a consistent ordering on the keys
- Output
  - “sorted” array A such that
    - For any i and j, if $i < j$ then $A[i] \le A[j]$

Consistent Ordering

- The comparison function must provide a ***consistent ordering*** on the set of possible keys
  - You can compare any two keys and get back an indication of a < b, a > b, or a = b (trichotomy)
  - The comparison functions must be consistent
    - If **`compare(a,b)`** says a<b, then **`compare(b,a)`** must say b>a
    - If **`compare(a,b)`** says a=b, then **`compare(b,a)`** must say b=a
    - If **`compare(a,b)`** says a=b, then **`equals(a,b) and equals(b,a)`** must say a=b

Why Sort?

- Provides fast search: binary $O(\log n)$
- Find *k*th largest element in: $O(1)$

Space

- How much space does the sorting algorithm require?
  - In-place: no more than the array or at most O(1) addition space
  - out-of-place: use separate data structures, copy back
  - External memory sorting – data so large that does not fit in memory

Stability

A sorting algorithm is **stable** if:

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

Input		Unstable sort		Stable Sort	
Adams	1	Adams	1	Adams	1
Black	2	Smith	1	Smith	1
Brown	4	Washington	2	Black	2
Jackson	2	Jackson	2	Jackson	2
Jones	4	Black	2	Washington	2
Smith	1	White	3	White	3
Thompson	4	Wilson	3	Wilson	3
Washington	2	Thompson	4	Brown	4
White	3	Brown	4	Jones	4
Wilson	3	Jones	4	Thompson	4

[Sedgewick]

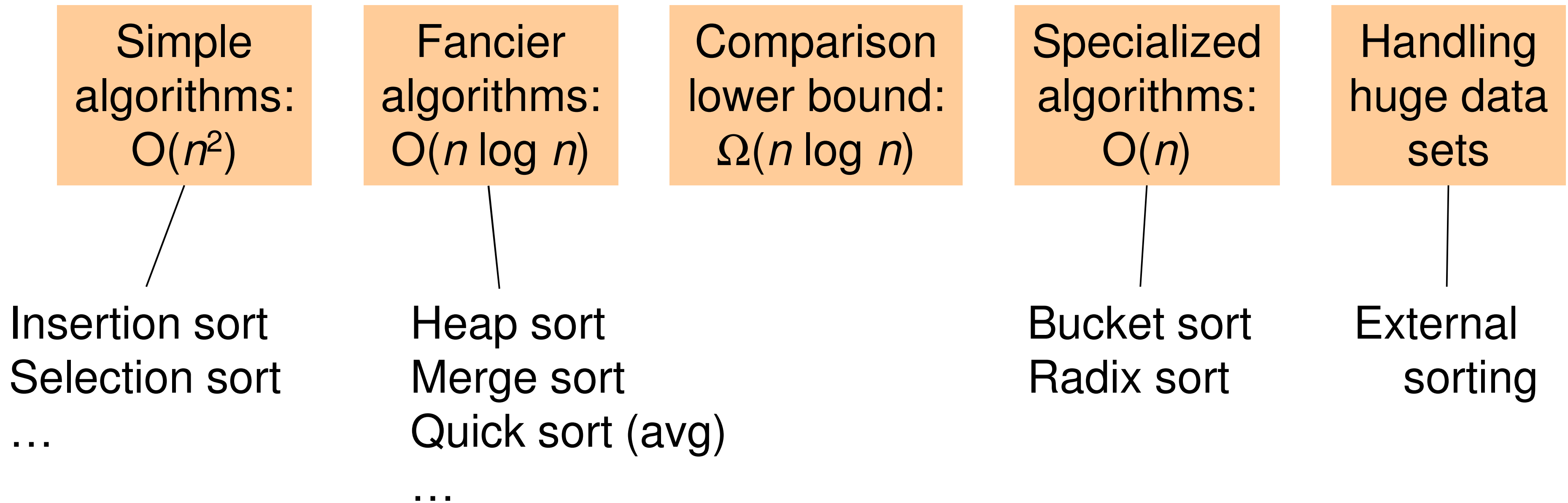
Time

How fast is the algorithm?

- requirement: for any i<j, A[i] ≤ A[j]
- This means that you need to at least check on each element at the very minimum
  - Complexity is at least: $O(n)$
- And you could end up checking each element against every other element
  - Complexity could be as bad as: $O(n^2)$

The big question: How close to $O(n)$ can you get?

Sorting: *The Big Picture*



Demo (with sound!)

- http://www.youtube.com/watch?v=kPRA0W1kECg

Selection Sort: idea

1. Find the smallest element, put it 1st
2. Find the next smallest element, put it 2nd
3. Find the next smallest, put it 3rd
4. And so on …

Try it out: Selection Sort

- 31, 16, 54, 4, 2, 17, 6

2 4 6 16 31 31 54

17

Selection Sort: Code

```
void SelectionSort (Array a[0..n-1]) {
    for (i=0; i<n; ++i) {
        j = Find index of
                smallest entry in a[i..n-1]
        Swap(a[i],a[j])
    }
}
```

O(n) swaps

Runtime:

worst case : $O(n^2)$

best case : $O(n^2)$

average case : $O(n^2)$

Bubble Sort

- Take a pass through the array
  - If neighboring elements are out of order, swap them.
- Repeat until no swaps needed

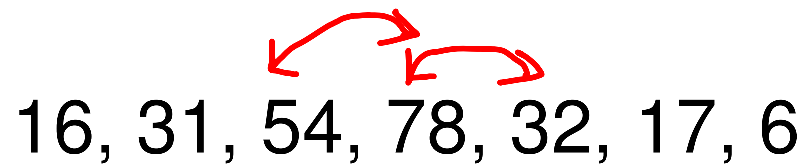
- Wost & avg case: $O(n^2)$
  - pretty much no reason to ever use this algorithm

Insertion Sort

1. Sort first 2 elements.
2. Insert 3rd element in order.
   - (First 3 elements are now sorted.)
3. Insert 4th element in order
   - (First 4 elements are now sorted.)
4. And so on…

How to do the insertion?

Suppose my sequence is:



16, 31, 54, 78, 32, 17, 6

And I've already sorted up to 78. How to insert 32?

Try it out: Insertion sort

- 31, 16, 54, 4, 2, 17, 6

Insertion Sort: Code

```
void InsertionSort (Array a[0..n-1]) {
    for (i=1; i<n; i++) {
        for (j=i; j>0; j--) {
            if (a[j] < a[j-1])
                Swap(a[j],a[j-1])
            else
                break
        }

}
```

Note: can instead move the "hole" to minimize copying, as with a binary heap.

$\text{avg} \sim \frac{1}{2} \sum_{1}^{n} i = O(n^2)$

Runtime:

worst case : $O(n^2)$

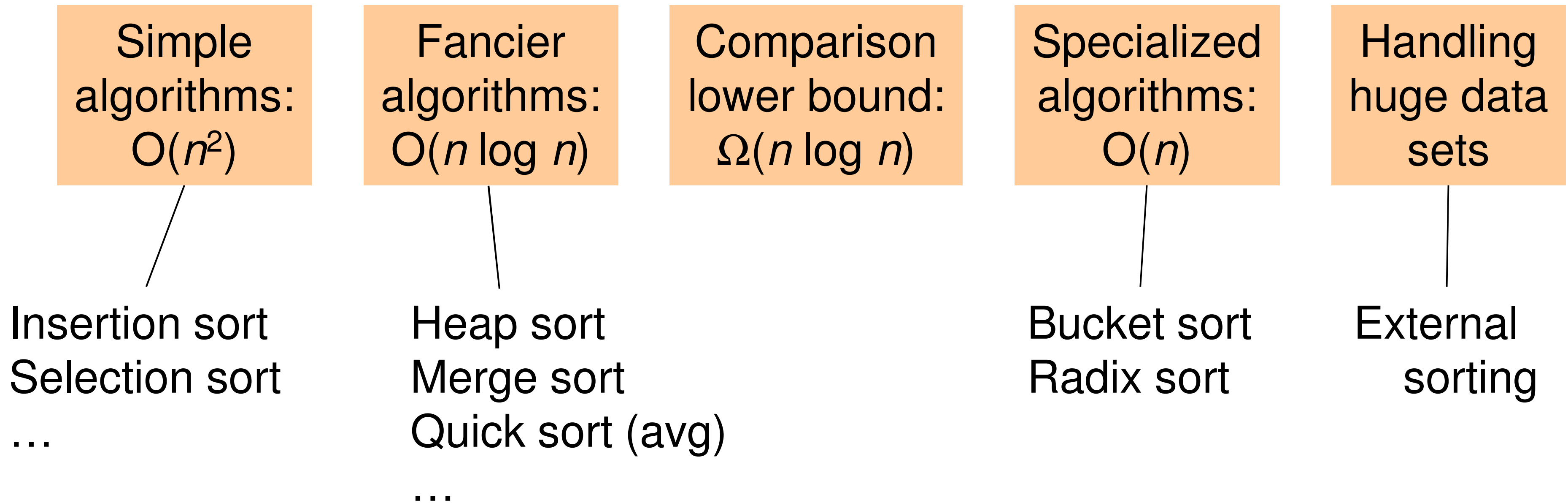
best case : $O(n)$

average case : $O(n^2)$

Insertion Sort vs. Selection Sort

- Same worst case, avg case complexity
- Insertion better best-case
  - preferable when input is "almost sorted"
    - one of the best sorting algs for almost sorted case (also for small arrays)

Sorting: *The Big Picture*



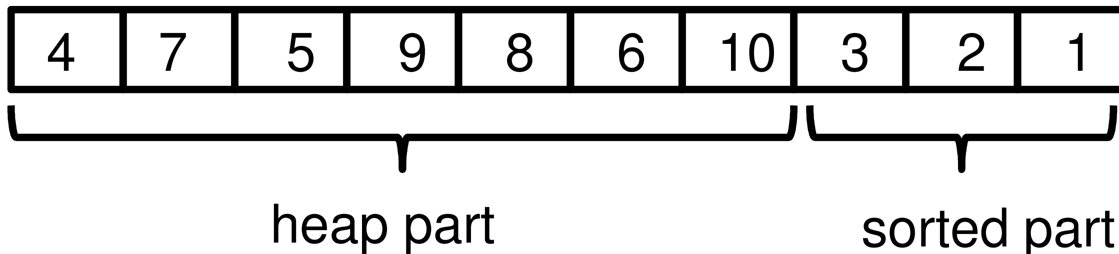
Heap Sort: Sort with a Binary Heap

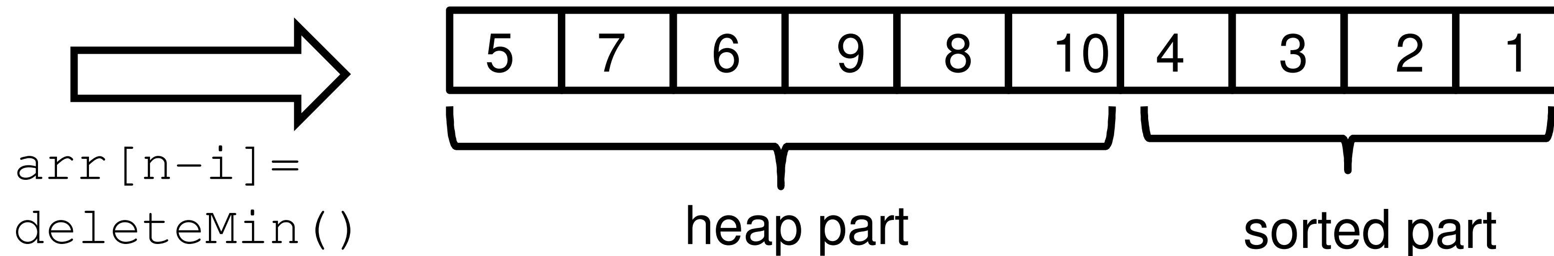
1. Build heap $O(n)$
2. deletemin n times $O(n \log n)$

Worst Case Runtime: $O(n \log n)$

In-place heap sort

- Treat the initial array as a heap (via **buildHeap**)
- When you delete the **i**th element, put it at **arr[n-i]**
  - It's not part of the heap anymore!





AVL Sort

1. build AVL tree (n inserts): $O(n \log n)$
2. in-order traversal: $O(n)$

Worst Case Runtime: $O(n \log n)$

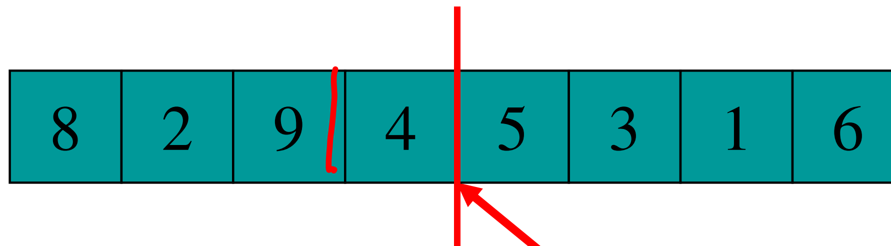
“Divide and Conquer”

- Very important strategy in computer science:
  - Divide problem into smaller parts
  - Independently solve the parts
  - Combine these solutions to get overall solution

- **Idea 1**: Divide array in half, *recursively* sort left and right halves, then *merge* two halves
  → known as Mergesort

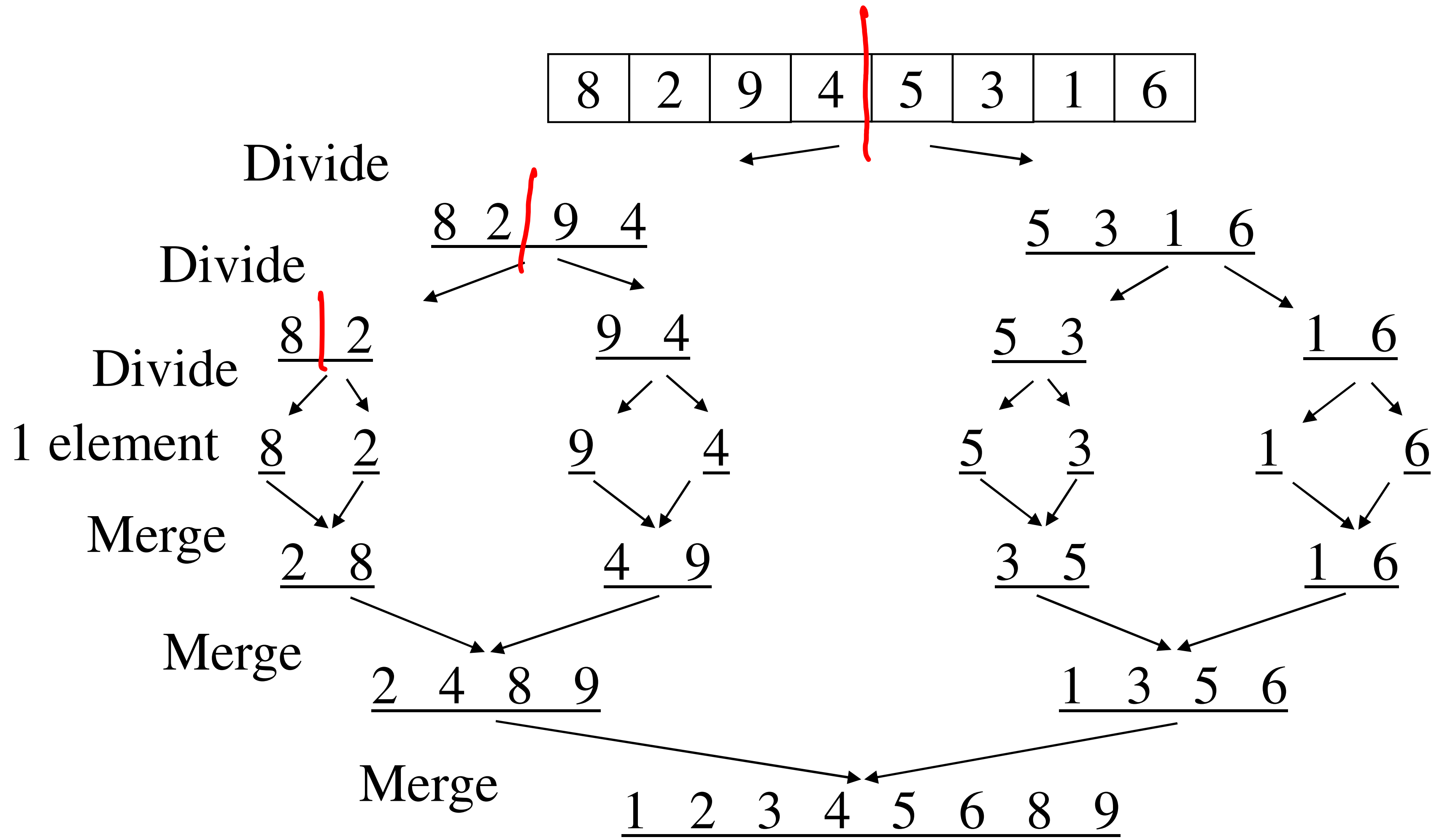
- **Idea 2 :** Partition array into small items and large items, then recursively sort the two sets
  → known as Quicksort

Mergesort



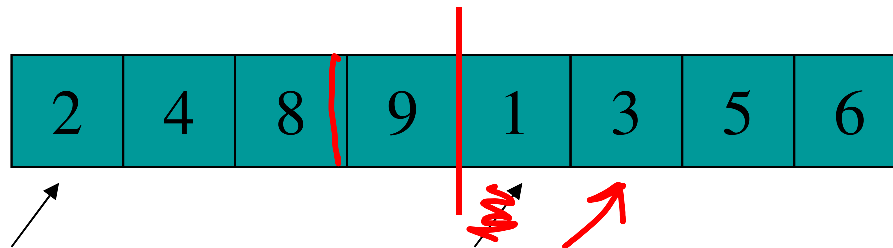
- Divide it in two at the midpoint
- Sort each half (recursively)
- Merge two halves together

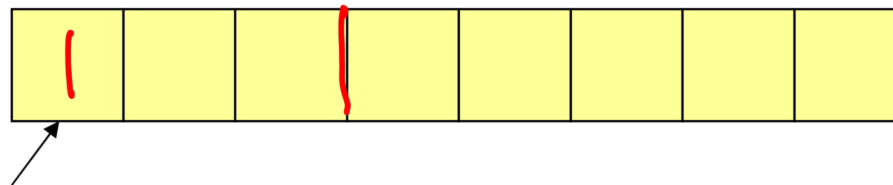
Mergesort Example



Merging: Two Pointer Method

- Perform merge using an auxiliary array

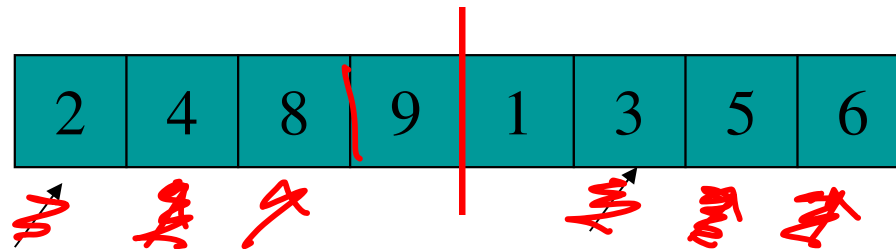


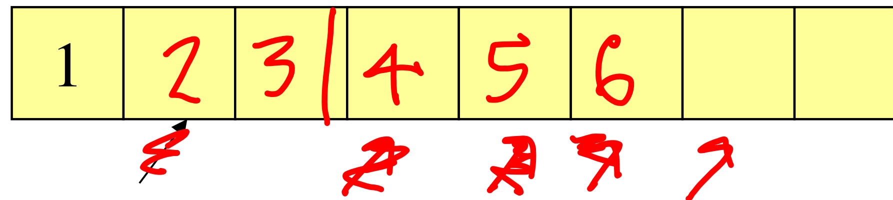


Auxiliary array

Merging: Two Pointer Method

- Perform merge using an auxiliary array

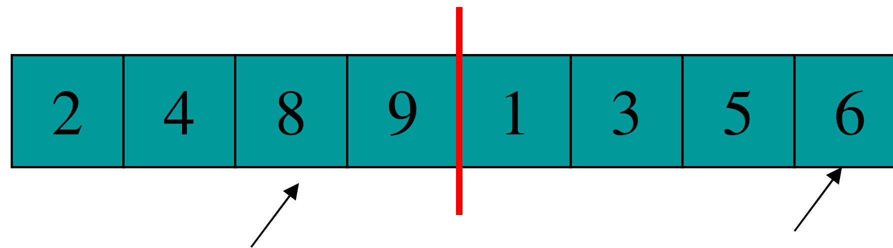


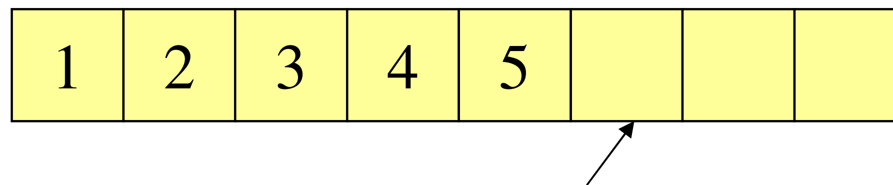


Auxiliary array

Merging: Two Pointer Method

- Perform merge using an auxiliary array

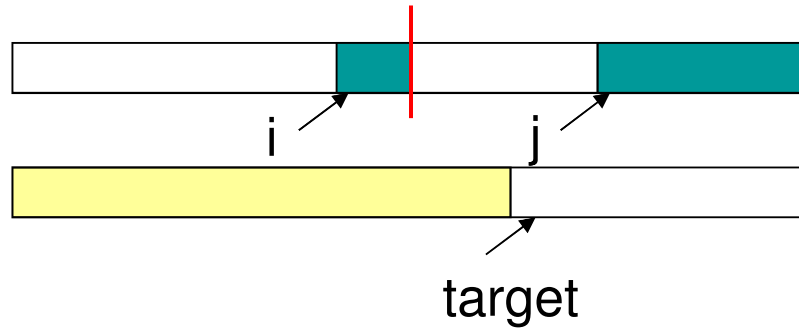




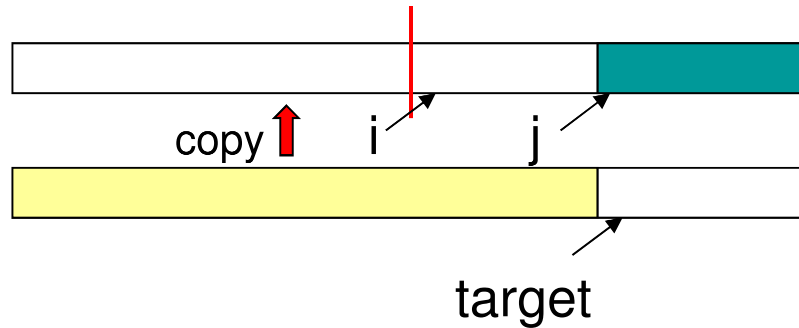
Auxiliary array

Merging: Finishing Up

Starting from here…

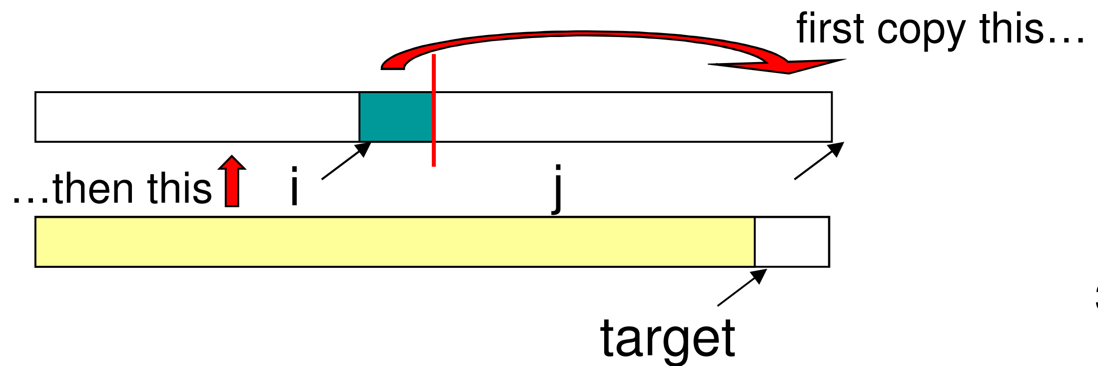


Left finishes up



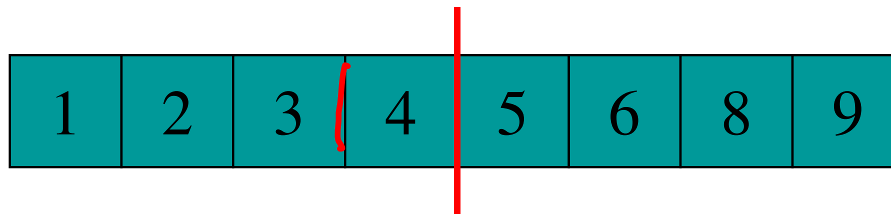
or

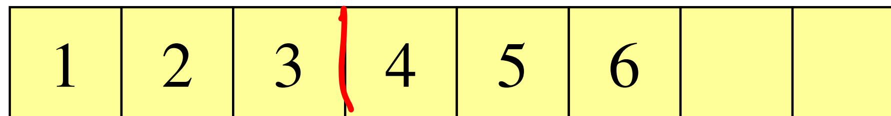
Right finishes up



Merging: Two Pointer Method

- Final result





Auxiliary array

Complexity? O(n)

Stability? yes if left wins on ties (= keys)

Merging

```
Merge(A[], Temp[], left, mid, right)  {
  Int i, j, k, l, target
  i = left
  j = mid + 1
  target = left
  while (i ≤ mid && j ≤ right) {
    if (A[i] ≤ A[j])
      Temp[target] = A[i++]
    else
      Temp[target] = A[j++]
    target++
  }
  if (i > mid) //left completed//
    for (k = left to target-1)
      A[k] = Temp[k];
  if (j > right) //right completed//
    k = mid
    l = right
    while (k ≥ i)
      A[l--] = A[k--]
    for (k = left to target-1)
      A[k] = Temp[k]
}
```

stability

Recursive Mergesort

```
MainMergesort(A[1..n], n) {
  Array Temp[1..n]
  Mergesort[A, Temp, 1, n]
}

Mergesort(A[], Temp[], left, right) {
  if (left < right) {
    mid = (left + right)/2
    Mergesort(A, Temp, left, mid)
    Mergesort(A, Temp, mid+1, right)
    Merge(A, Temp, left, mid, right)
  }
}
```

What is the recurrence relation?

$T(1) = 1$

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

Mergesort: Complexity

$T(1) = 1 \quad T(n) = 2T(n/2) + n$

$= 2(2T(n/4) + n/2) + n$

$= 2(2(2T(n/8) + n/4) + n/2) + n$

$= 8T(n/8) + 4\,n/4 + 2\,n/2 + n$

$= 8T(n/8) + n + n + n$

$= 8T(n/8) + 3n$

$= 2^3 T(n/2^3) + 3n$

$= 2^k T(n/2^k) + kn$

$= n \cdot 1 + (\log_2 n)n$

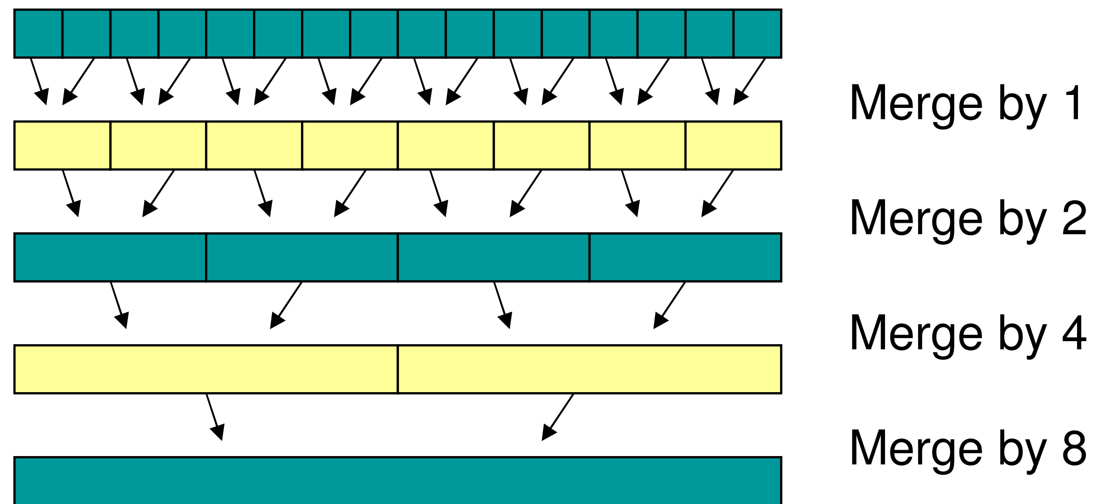
$\in O(n \log n)$

$n/2^k = 1$

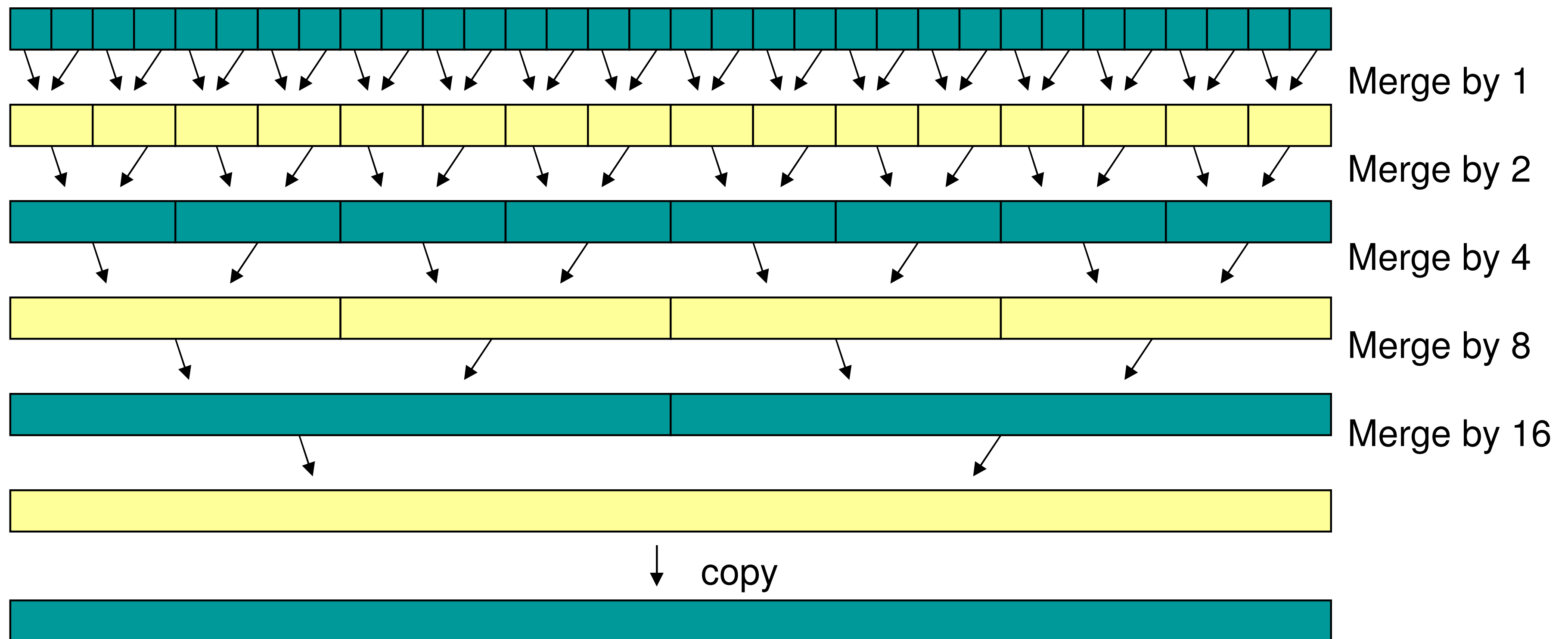
$n = 2^k$

$k = \log_2 n$

Iterative Mergesort



Iterative Mergesort



Iterative Mergesort reduces copying

Complexity? O(n log n)

Properties of Mergesort

- In-place? no
- Stable? yes
- Sorted list complexity? $O(n \log n)$
- Nicely extends to handle linked lists.
- Multi-way merge is basis of big data sorting.
- Java uses Mergesort on Collections and on Arrays of Objects.

Quicksort

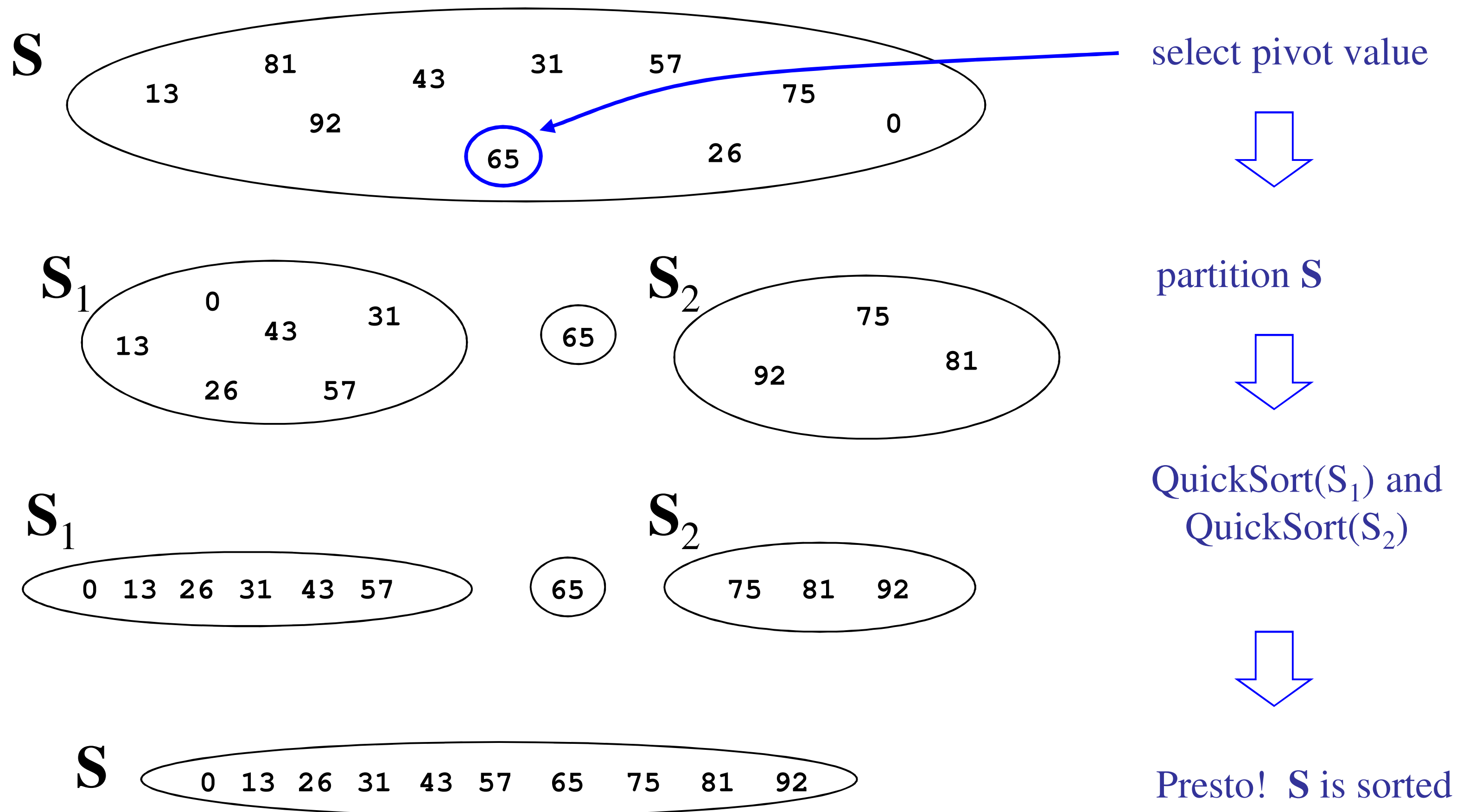
Quicksort uses a divide and conquer strategy, but does not require the O(N) extra space that MergeSort does.

Here's the idea for sorting array **S**:

1. Pick an element *v* in **S**. This is the ***pivot*** value.
2. Partition **S**-{*v*} into two disjoint subsets, $\mathbf{S}_1$ and $\mathbf{S}_2$ such that:
   - elements in $\mathbf{S}_1$ are all $\leq v$
   - elements in $\mathbf{S}_2$ are all $\geq v$
3. Return concatenation of QuickSort($\mathbf{S}_1$), *v*, QuickSort($\mathbf{S}_2$)

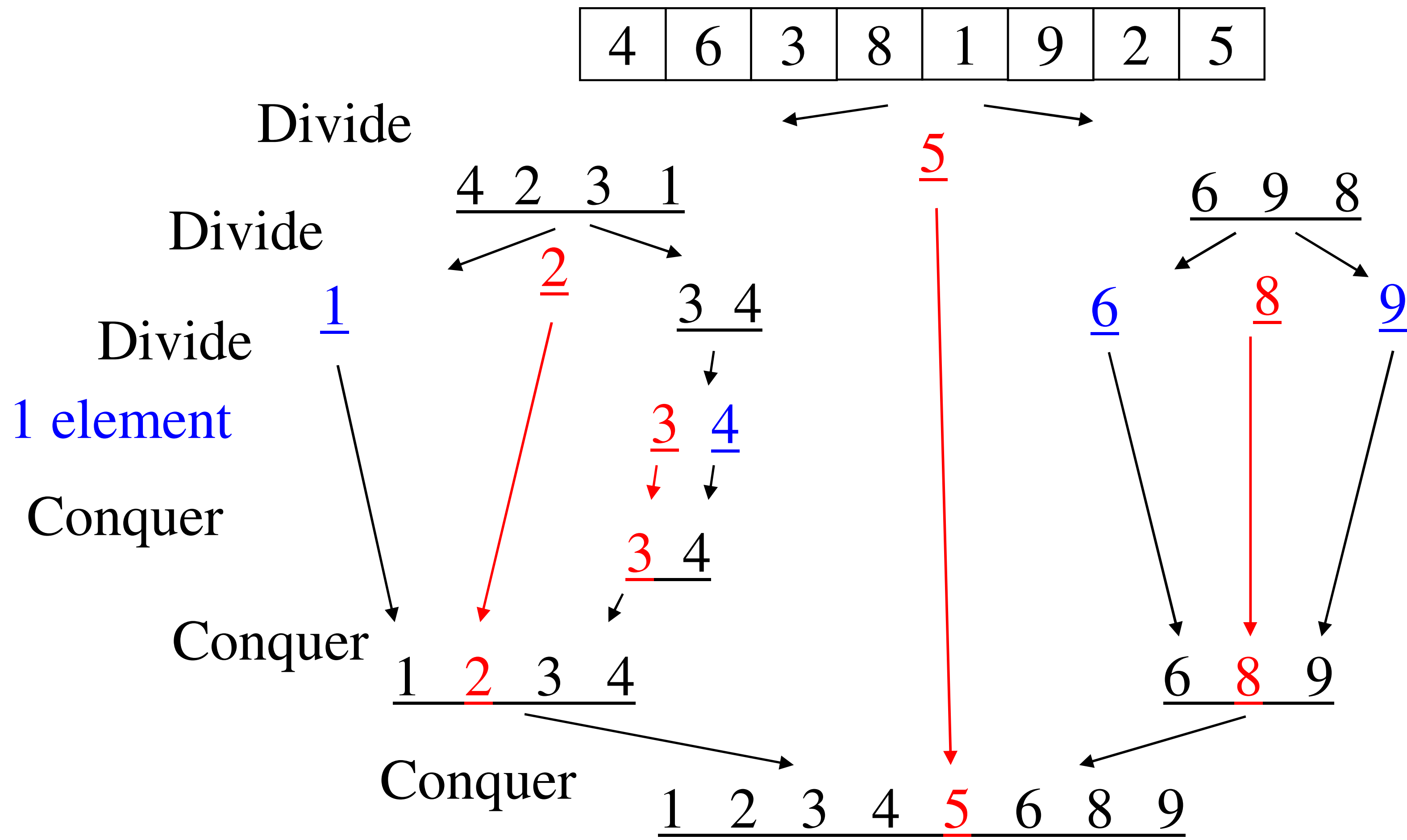
Recursion ends when Quicksort( ) receives an array of length 0 or 1.

The steps of Quicksort



[Weiss]

Quicksort Example



Pivot Picking and Partitioning

The tricky parts are:

- **Picking the pivot**
  - Goal: pick a pivot value so that $|S_1|$ and $|S_2|$ are roughly equal in size.
- **Partitioning**
  - Preferably in-place
  - Dealing with duplicates

Picking the Pivot

median is ideal - too expensive

middle - could be really bad

median of 3 - good

~ more than 3, hierarchical

Median of Three Pivot

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

`medianOf3Pivot(…)`

0	1	4	9	7	3	5	2	6	8

Choose the pivot as the median of three.

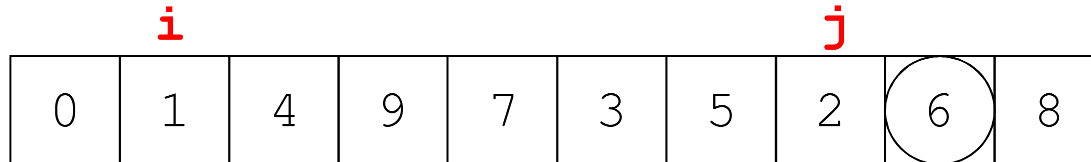
Place the pivot and the largest at the right and the smallest at the left.

Quicksort Partitioning

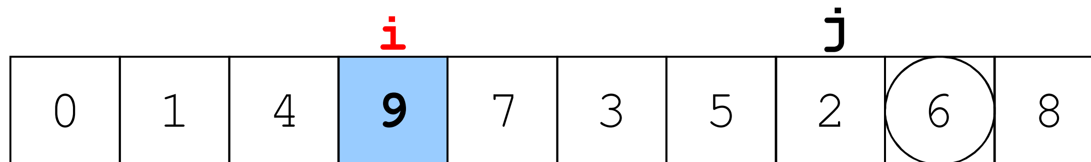
- Partition the array into left and right sub-arrays such that:
  - elements in left sub-array are ≤ pivot
  - elements in right sub-array are ≥ pivot
- Can be done in-place with another “two pointer method”
  - Sounds like mergesort, but here we are *partitioning*, not sorting…
  - …and we can do it in-place.

Partioning In-place

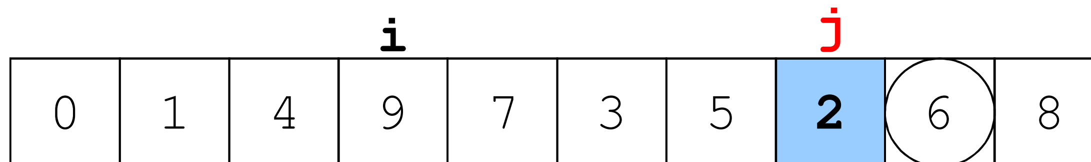
Setup: `i` = start and `j` = end of un-partioned elements:



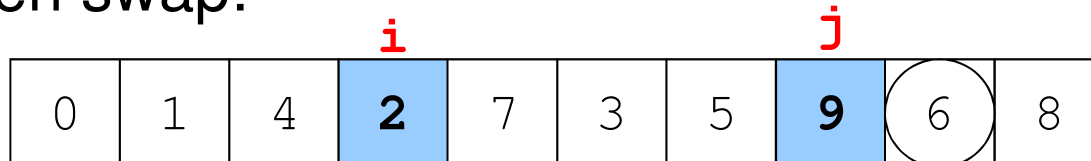
Advance `i` until element ≥ pivot:



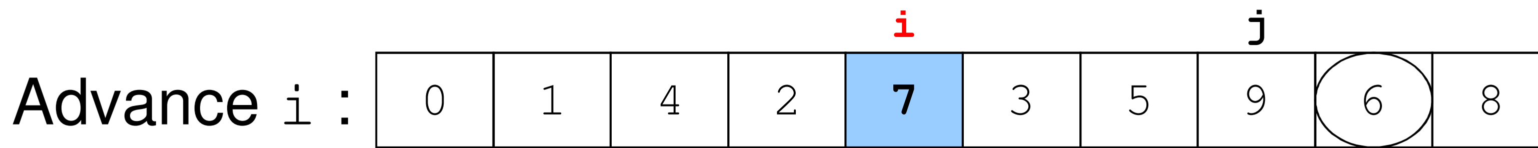
Advance `j` until element ≤ pivot:

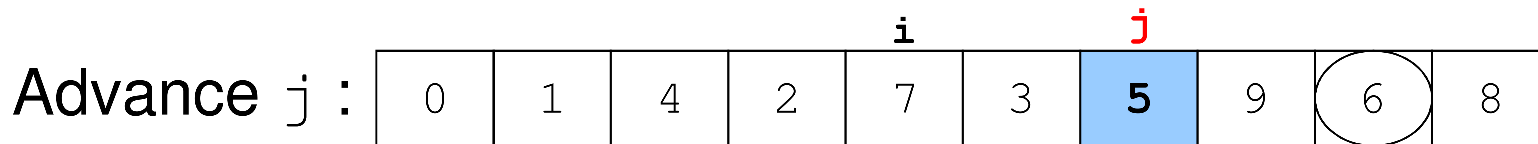


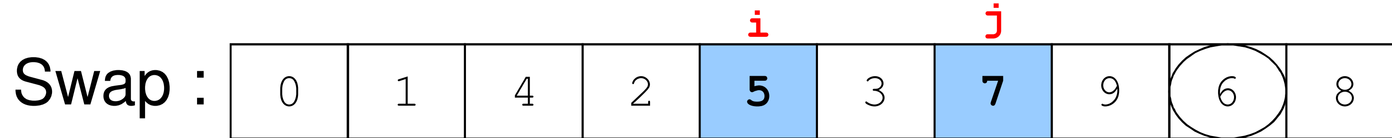
If `j` > `i`, then swap:

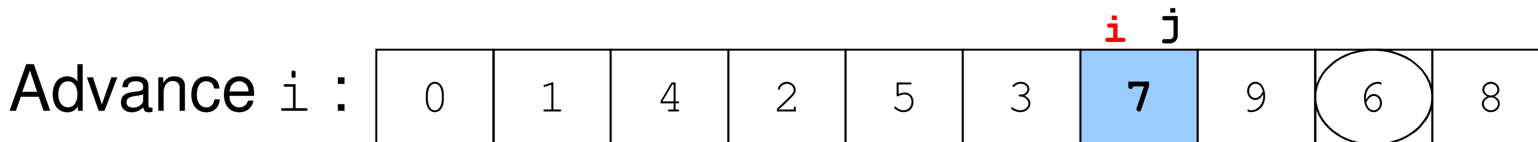


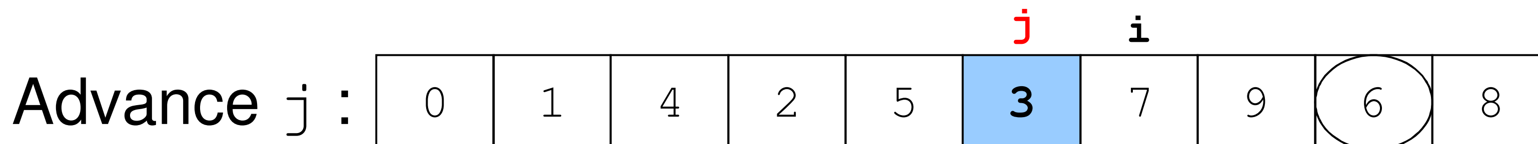
Partioning In-place



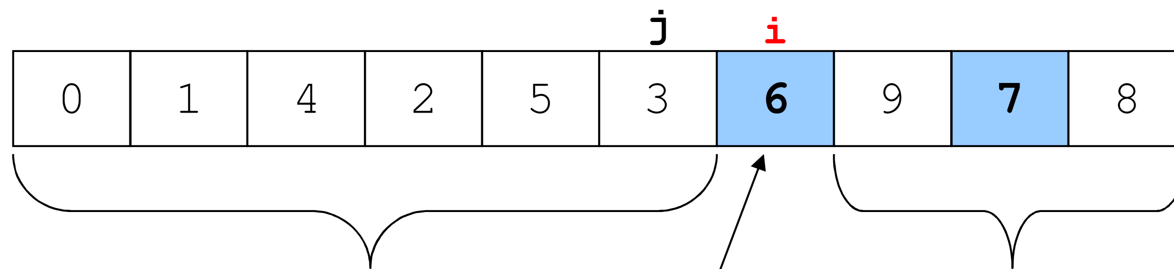








i > j, swap in pivot, partition done!



Partition Pseudocode

```
Partition(A[], left, right) {
  v = A[right]; // Assumes pivot value currently at right
  i = left;     // Initialize left side, right side pointers
  j = right-1;

  // Do i++, j-- until they cross, swapping values as needed
  while (1) {
    while (A[i] < v) i++;
    while (A[j] > v) j--;
    if (i < j) {
      Swap(A[i], A[j]);
      i++; j--;
    }
    else
      break;
  }
  Swap(A[i], A[right]); // Swap pivot value into position
  return i;             // Return the final pivot position
}
```

Complexity for input size *n*?

Quicksort Pseudocode

Putting the pieces together:

```
Quicksort(A[], left, right) {
  if (left < right) {
    medianOf3Pivot(A, left, right);
    pivotIndex = Partition(A, left+1, right-1);

    Quicksort(A, left, pivotIndex - 1);
    Quicksort(A, pivotIndex + 1, right);
  }
}
```

QuickSort: <u>Best</u> case complexity

```
Quicksort(A[], left, right) {
  if (left < right) {
    medianOf3Pivot(A, left, right);
    pivotIndex = Partition(A, left+1, right-1);

    Quicksort(A, left, pivotIndex - 1);
    Quicksort(A, pivotIndex + 1, right);
  }
}
```

QuickSort: Worst case complexity

```
Quicksort(A[], left, right) {
  if (left < right) {
    medianOf3Pivot(A, left, right);
    pivotIndex = Partition(A, left+1, right-1);

    Quicksort(A, left, pivotIndex - 1);
    Quicksort(A, pivotIndex + 1, right);
  }
}
```

QuickSort: Average case complexity

Turns out to be **O(n log n)**.

See Section 7.7.5 for an idea of the proof.
Don’t need to know proof details for this course.

Many Duplicates?

An important case to consider is when an array has many duplicates.

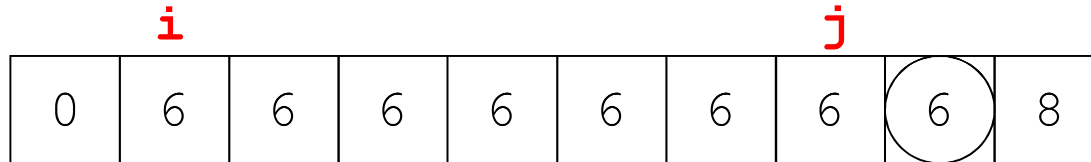
0	1	2	3	4	5	6	7	8	9
8	6	6	6	**6**	6	6	6	6	**0**

`medianOf3Pivot(…)`

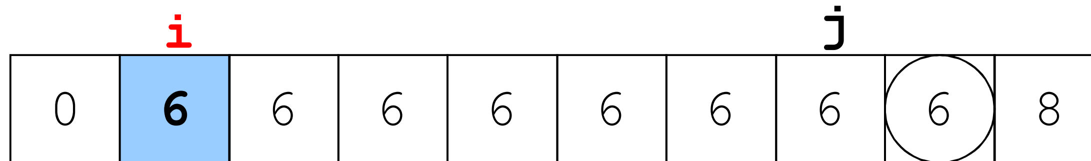
0	6	6	6	6	6	6	6	6	8

Partitioning with Duplicates

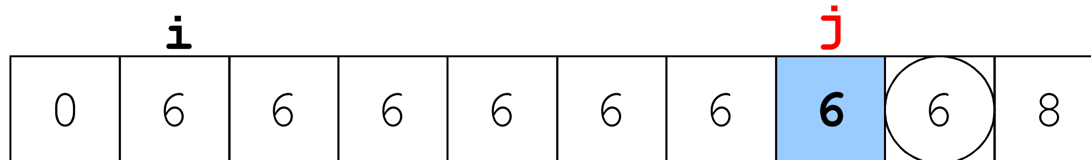
Setup: `i` = start and `j` = end of un-partioned elements:



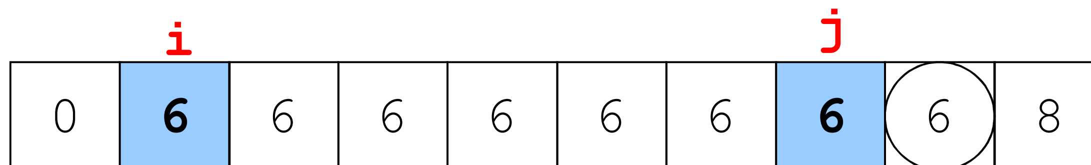
Advance `i` until element ≥ pivot:



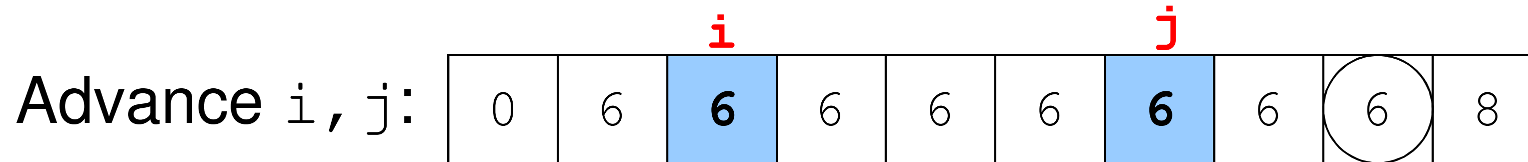
Advance `j` until element ≤ pivot:

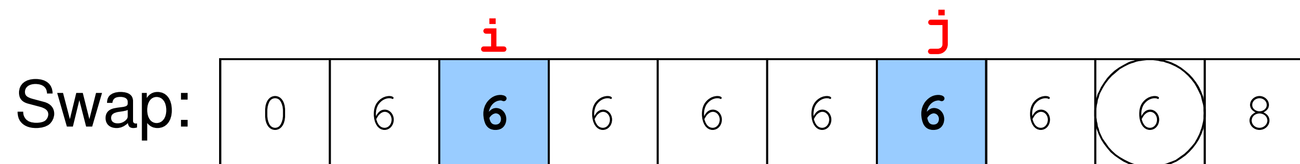


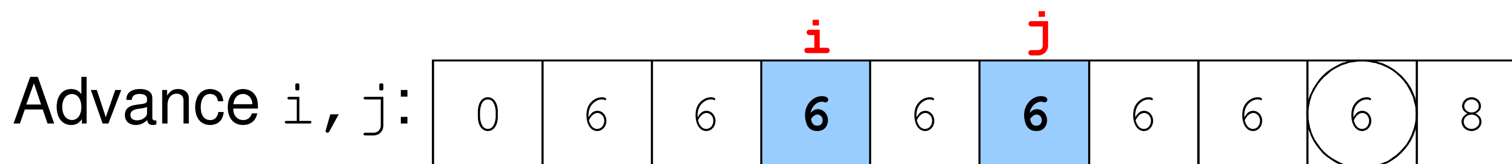
If `j` > `i`, then swap:

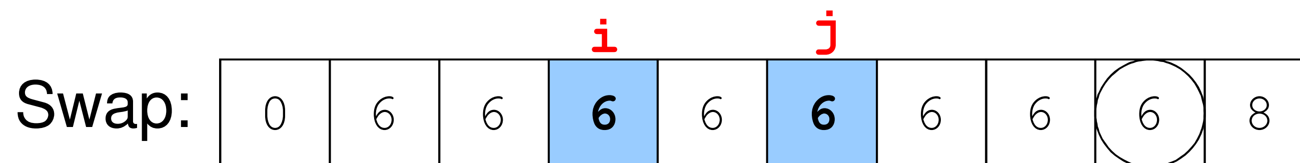


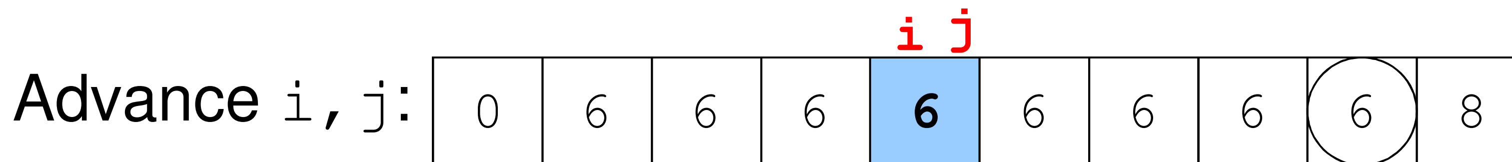
Partitioning with Duplicates

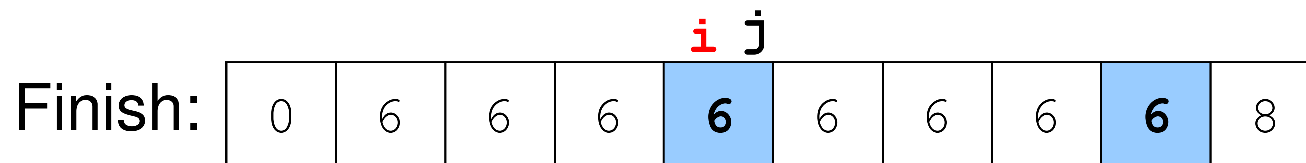






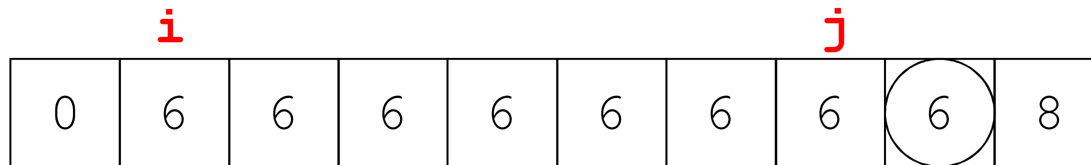




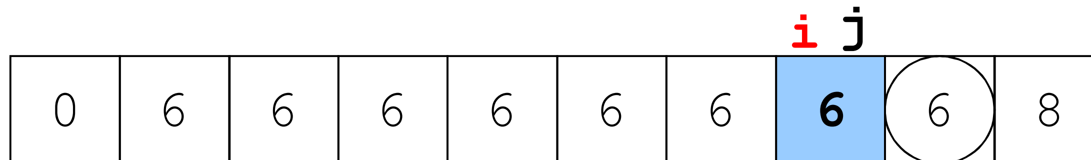


Partitioning with Duplicates:Take Two

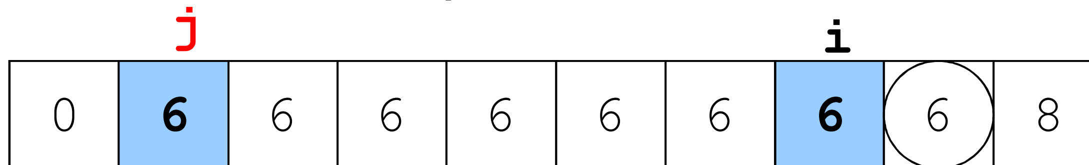
Start `i` = start and `j` = end of un-partioned elements:

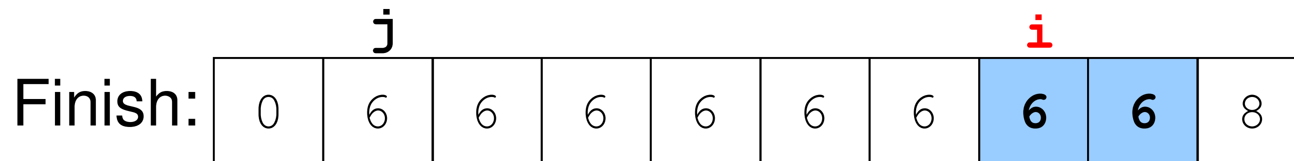


Advance `i` until element **>** pivot (and in bounds):



Advance `j` until element **<** pivot (and in bounds):





Is this better?

Partitioning with Duplicates: Upshot

It's better to stop advancing pointers when elements are equal to pivot, and then just do swaps.

Complexity of quicksort on an array of identical values?

Can we do better?

Important Tweak

Insertion sort is actually better than quicksort on small arrays. Thus, a better version of quicksort:

```
Quicksort(A[], left, right) {
  if (right - left ≥ CUTOFF) {
    medianOf3Pivot(A, left, right);
    pivotIndex = Partition(A, left+1, right-1);

    Quicksort(A, left, pivotIndex - 1);
    Quicksort(A, pivotIndex + 1, right);

  } else {
    InsertionSort(A, left, right);
  }
}
```

CUTOFF = 10 is reasonable.

Properties of Quicksort

- $O(N^2)$ worst case performance, but
  $O(N \log N)$ average case performance.
- Pure quicksort not good for small arrays.
- No iterative version (without using a stack).
- “In-place,” but uses auxiliary storage because of recursive calls.
- Stable?
- Used by Java for sorting arrays of primitive types.