



Sorting and Asymptotic Complexity

Lecture 13
CS2110 – Fall 2008

Prelim Tonight

InsertionSort

```
//sort a[], an array of int
for (int i = 1; i < a.length; i++) {
    int temp = a[i];
    int k;
    for (k = i; 0 < k && temp < a[k-1]; k--)
        a[k] = a[k-1];
    a[k] = temp;
}
```

- Many people sort cards this way
- Invariant: everything to left of i is already sorted
- Works especially well when input is *nearly sorted*
- Worst-case is $O(n^2)$
 - Consider reverse-sorted input
- Best-case is $O(n)$
 - Consider sorted input
- Expected case is $O(n^2)$
 - Expected number of inversions is $n(n-1)/4$

SelectionSort

- To sort an array of size n :
 - Examine $a[0]$ to $a[n-1]$; find the smallest one and swap it with $a[0]$
 - Examine $a[1]$ to $a[n-1]$; find the smallest one and swap it with $a[1]$
 - In general, in step i , examine $a[i]$ to $a[n-1]$; find the smallest one and swap it with $a[i]$
- This is the other common way for people to sort cards
- Runtime
 - Worst-case $O(n^2)$
 - Best-case $O(n^2)$
 - Expected-case $O(n^2)$

Divide & Conquer?

- It often pays to
 - Break the problem into smaller subproblems,
 - Solve the subproblems separately, and then
 - Assemble a final solution
- This technique is called *divide-and-conquer*
 - Caveat: It won't help unless the *partitioning* and *assembly* processes are inexpensive
- Can we apply this approach to sorting?

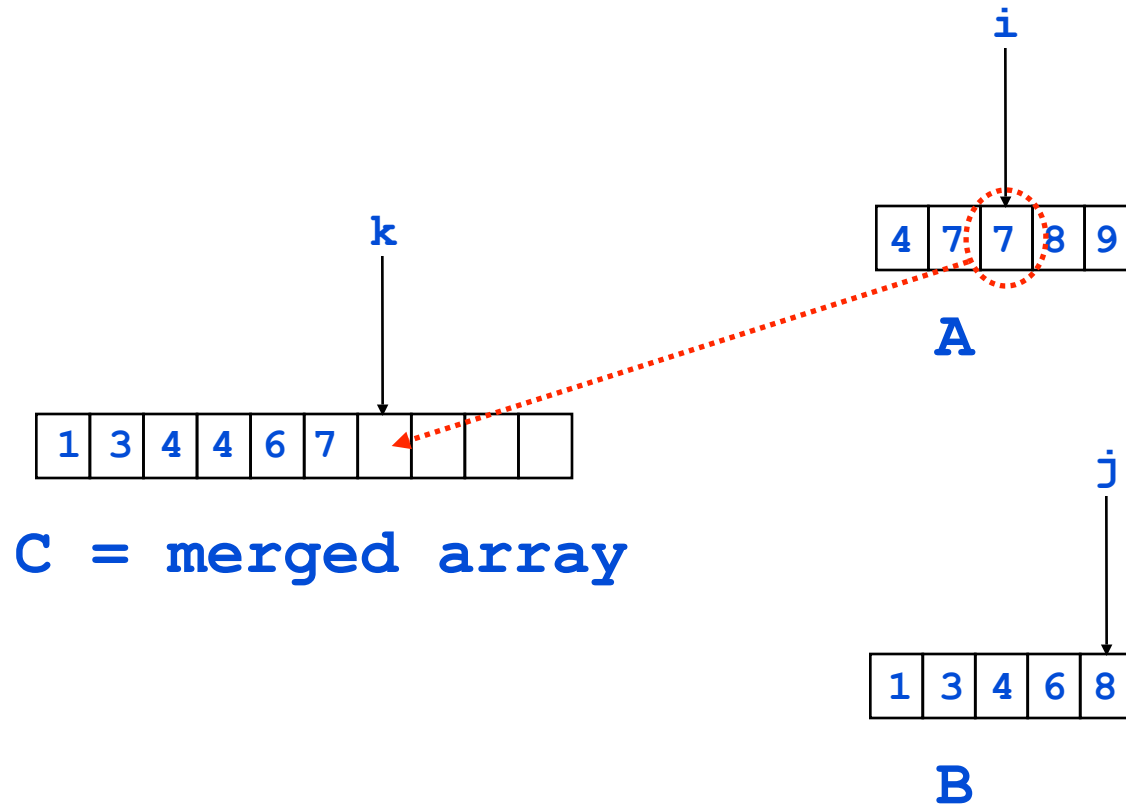
MergeSort

- Quintessential divide-and-conquer algorithm
- Divide array into equal parts, sort each part, then merge
- Questions:
 - Q1: How do we divide array into two equal parts?
 - A1: Find middle index: `a.length/2`
 - Q2: How do we sort the parts?
 - A2: call `MergeSort` recursively!
 - Q3: How do we merge the sorted subarrays?
 - A3: We have to write some (easy) code

Merging Sorted Arrays **A** and **B**

- Create an array **C** of size = size of **A** + size of **B**
- Keep three indices:
 - **i** into **A**
 - **j** into **B**
 - **k** into **C**
- Initialize all three indices to 0 (start of each array)
- Compare element **A[i]** with **B[j]**, and move the smaller element into **C[k]**
- Increment **i** or **j**, whichever one we took, and **k**
- When either **A** or **B** becomes empty, copy remaining elements from the other array (**B** or **A**, respectively) into **C**

Merging Sorted Arrays



MergeSort Analysis

- Outline (detailed code on the website)
 - Split array into two halves
 - Recursively sort each half
 - Merge the two halves
- Merge = combine two sorted arrays to make a single sorted array
 - Rule: always choose the smallest item
 - Time: $O(n)$ where n is the combined size of the two arrays
- Runtime recurrence
 - Let $T(n)$ be the time to sort an array of size n
 - $T(n) = 2T(n/2) + O(n)$
 - $T(1) = 1$
- Can show by induction that $T(n)$ is $O(n \log n)$
- Alternately, can see that $T(n)$ is $O(n \log n)$ by looking at tree of recursive calls

MergeSort Notes

- Asymptotic complexity: $O(n \log n)$
 - Much faster than $O(n^2)$
- Disadvantage
 - Need extra storage for temporary arrays
 - In practice, this can be a disadvantage, even though **MergeSort** is *asymptotically optimal for sorting*
 - Can do **MergeSort** in place, but this is *very tricky* (and it slows down the algorithm significantly)
- Are there good sorting algorithms that do not use so much extra storage?
 - Yes: **QuickSort**

QuickSort

- Intuitive idea
 - Given an array \mathbf{A} to sort, choose a pivot value p
 - Partition \mathbf{A} into two subarrays, \mathbf{AX} and \mathbf{AY}
 - ♦ \mathbf{AX} contains only elements $\leq p$
 - ♦ \mathbf{AY} contains only elements $\geq p$
 - Sort subarrays \mathbf{AX} and \mathbf{AY} separately
 - *Concatenate* (not merge!) sorted \mathbf{AX} and \mathbf{AY} to get sorted \mathbf{A}
 - ♦ Concatenation is easier than merging – $O(1)$

20	31	24	19	45	56	4	65	5	72	14	99
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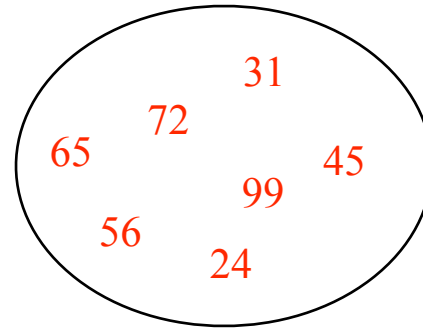
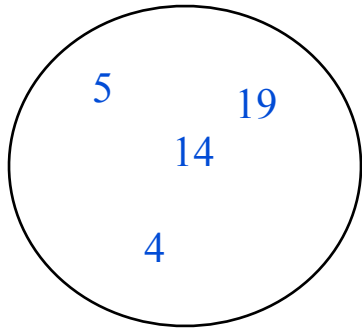
pivot

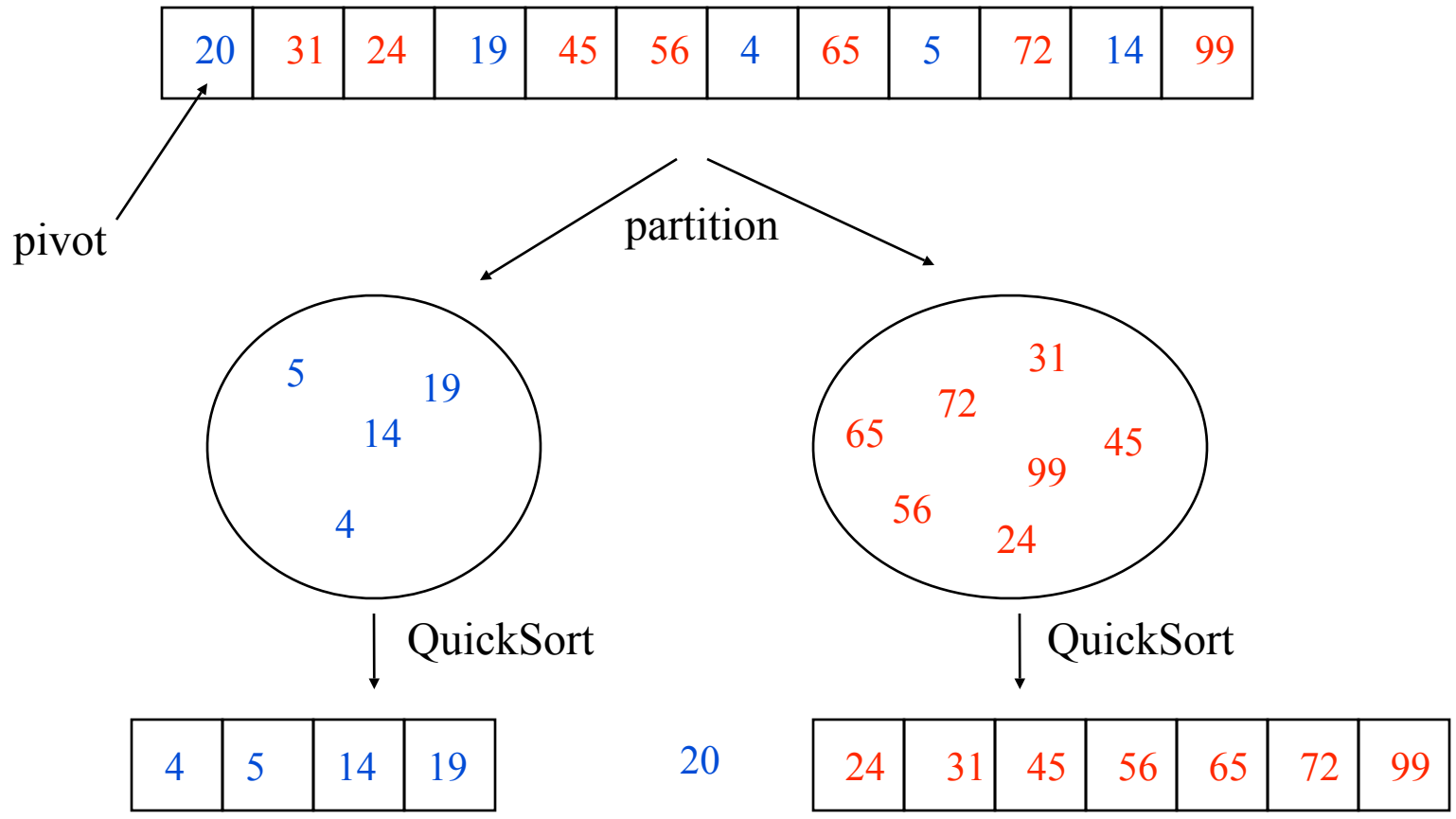


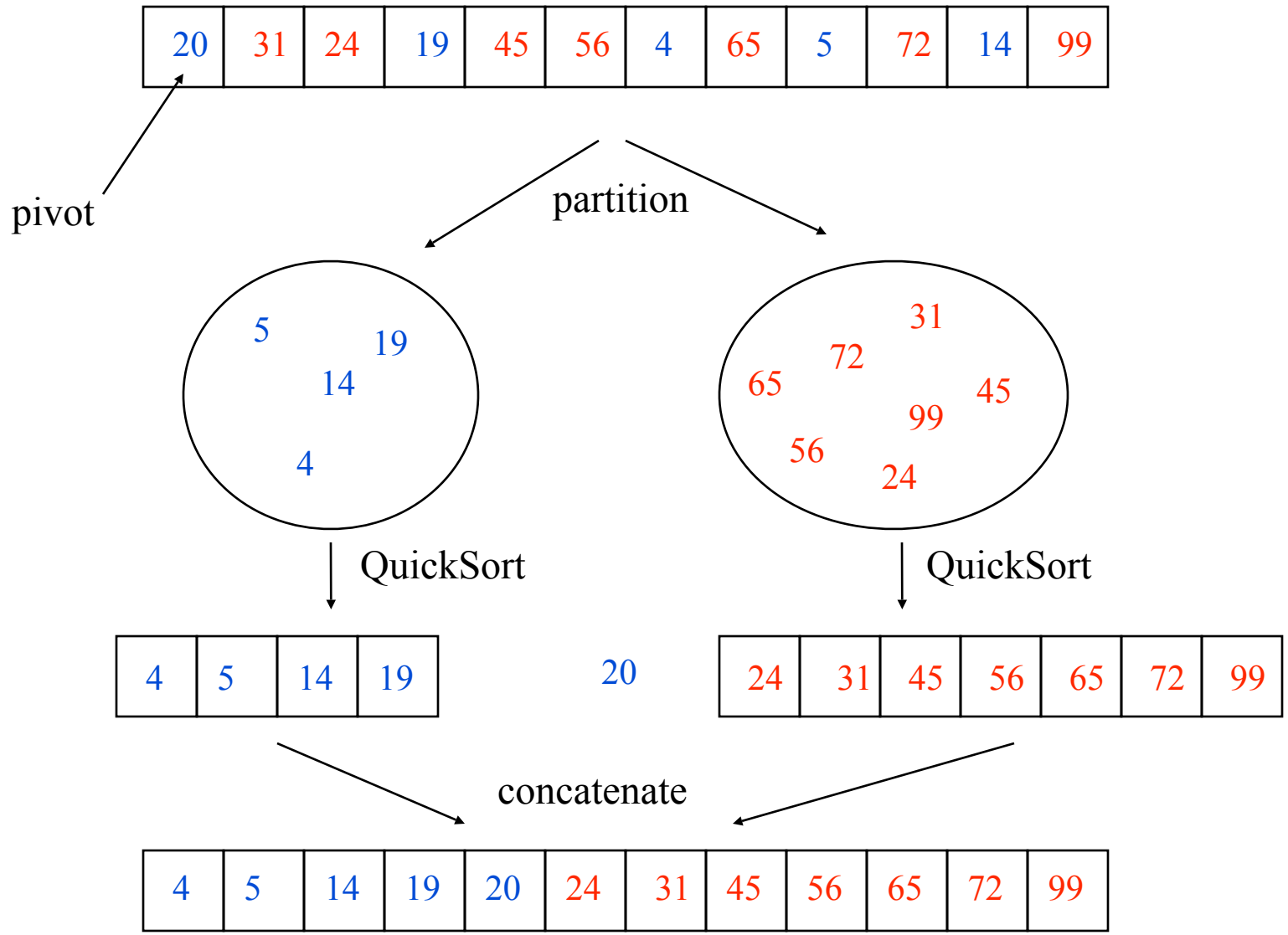


pivot

partition



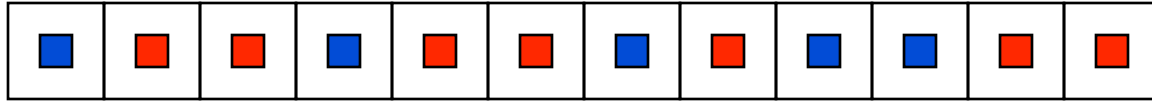




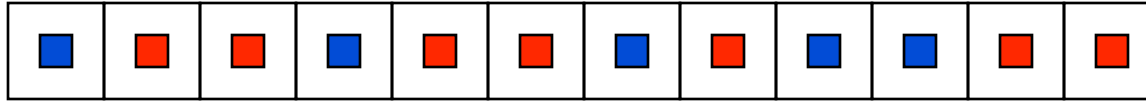
QuickSort Questions

- Key problems
 - How should we choose a *pivot*?
 - How do we *partition* an array in place?
- Partitioning in place
 - Can be done in $O(n)$ time (next slide)
- Choosing a pivot
 - Ideal pivot is the median, since this splits array in half
 - Computing the median of an unsorted array is $O(n)$, but algorithm is quite complicated
 - Popular heuristics:
 - ◆ Use first value in array (usually not a good choice)
 - ◆ Use middle value in array
 - ◆ Use median of first, last, and middle values in array
 - ◆ Choose a random element

In-Place Partitioning



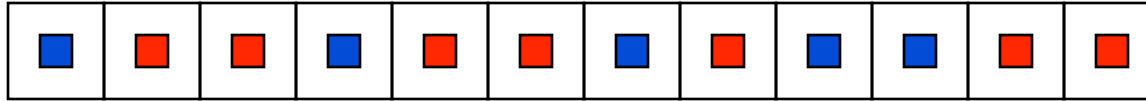
In-Place Partitioning



How can we move all the blues to the left of all the reds?

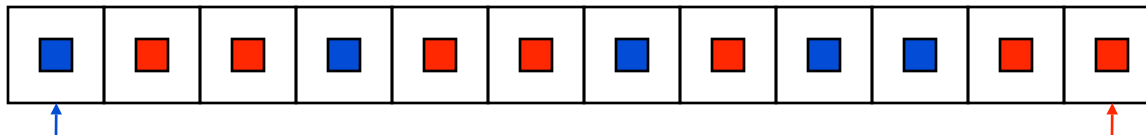
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2. Initialize LEFT at start of array and RIGHT at end of array
3. Invariant: all elements to left of LEFT are blue
all elements to right of RIGHT are red
4. Keep advancing indices until they pass, maintaining invariant

In-Place Partitioning

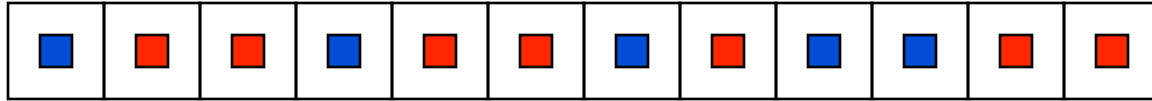


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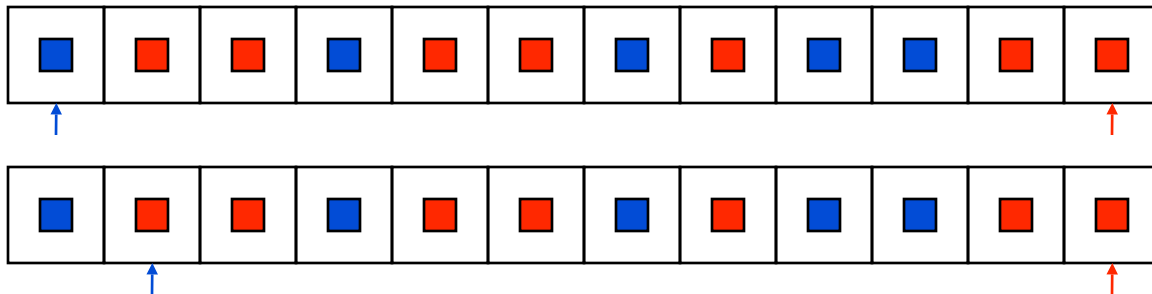


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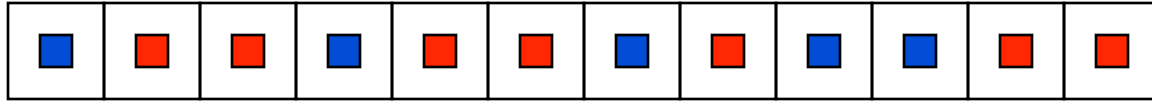


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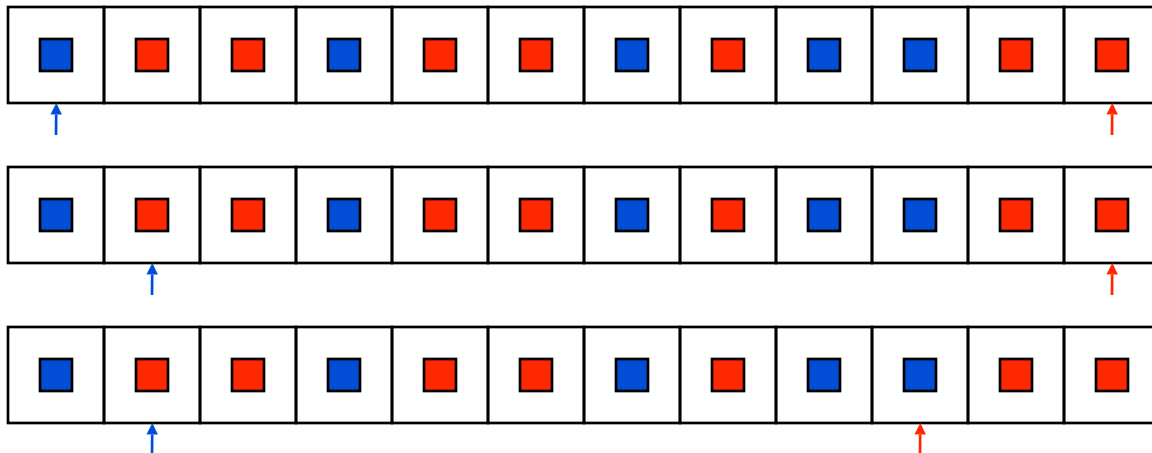


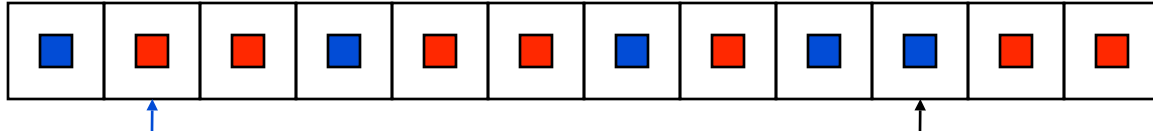
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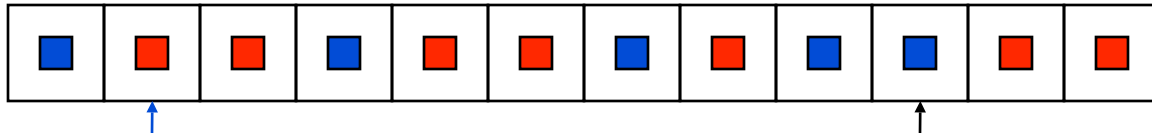
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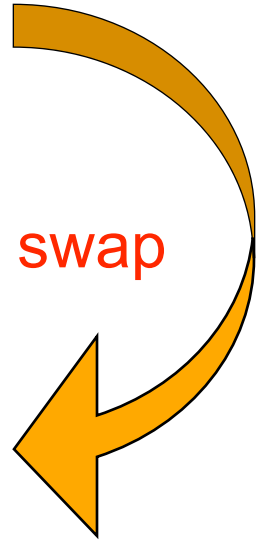
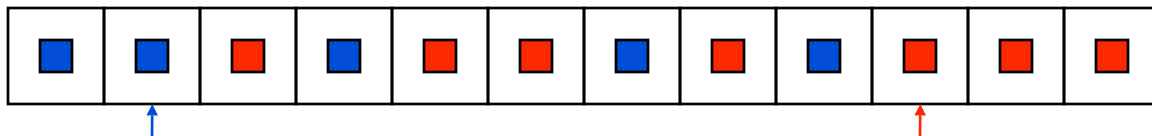


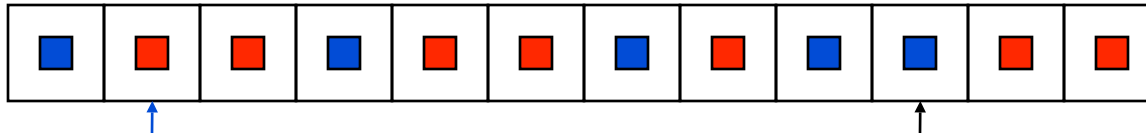


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We can swap red and blue pointed to by LEFT and RIGHT indices.
After swap, indices can continue to advance until next conflict.

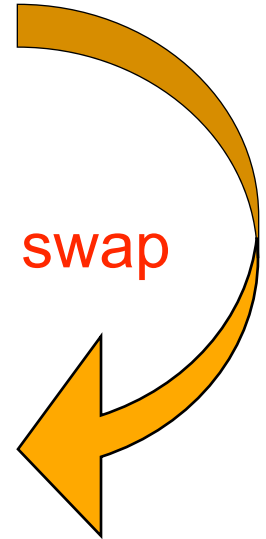
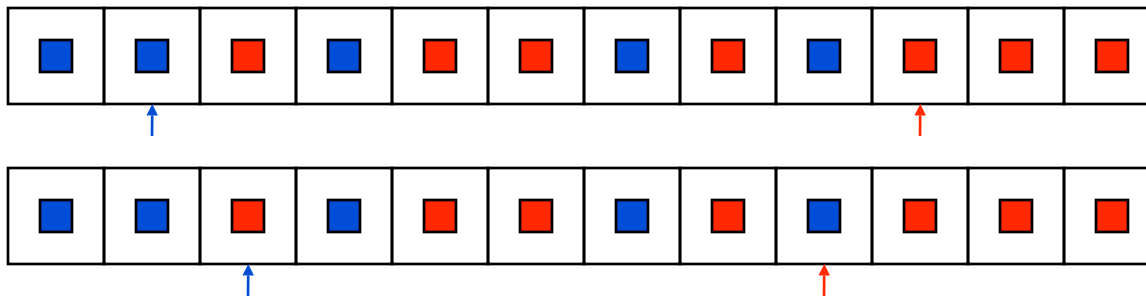


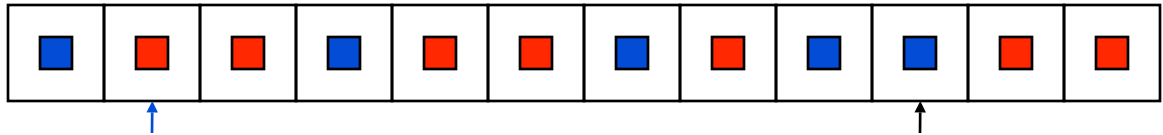
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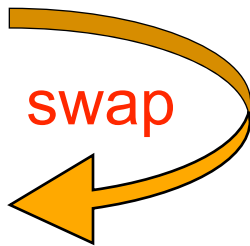
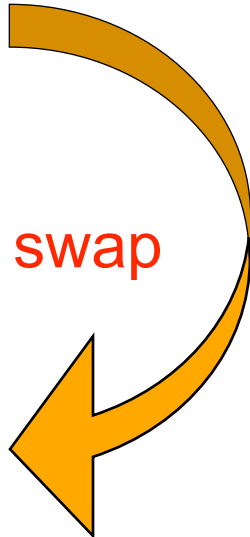
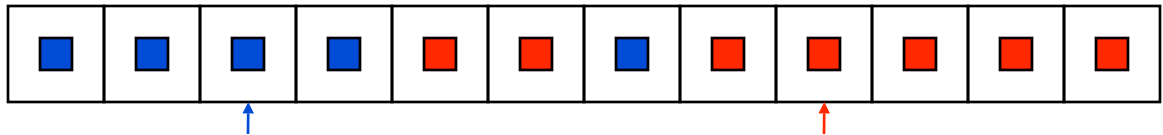
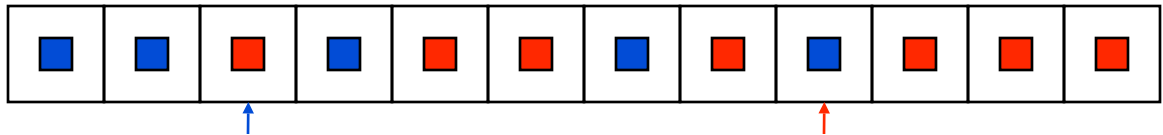
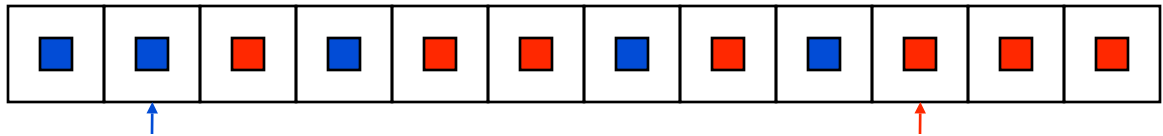


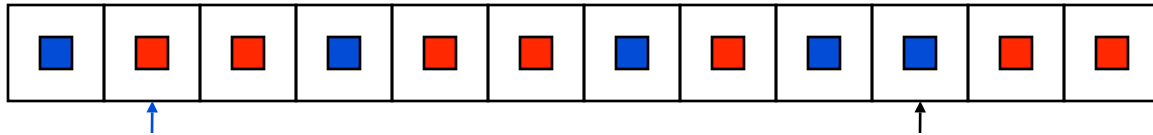
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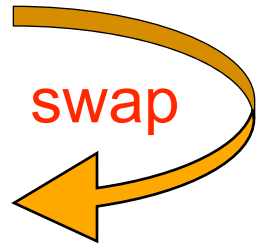
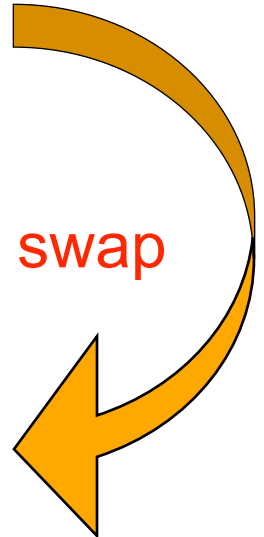
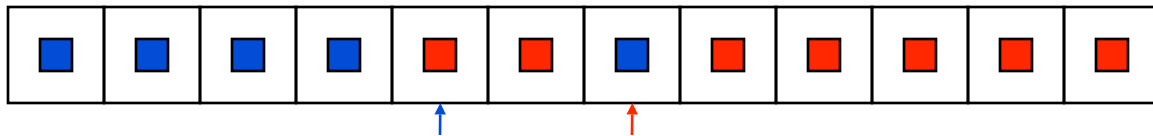
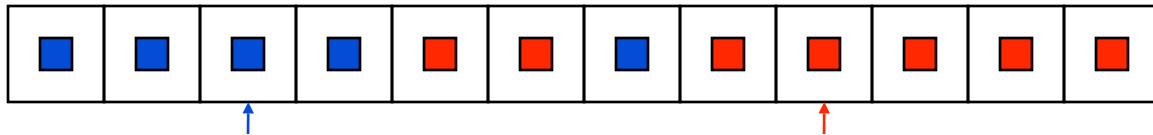
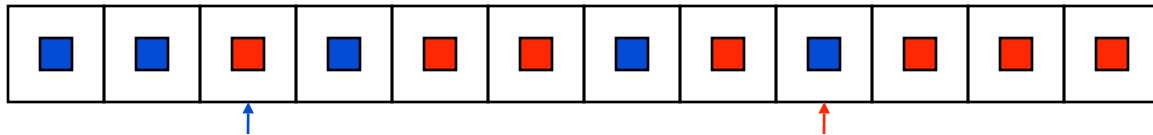
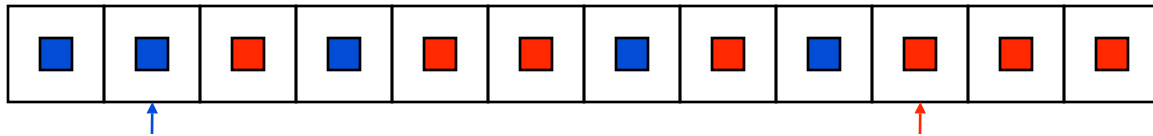


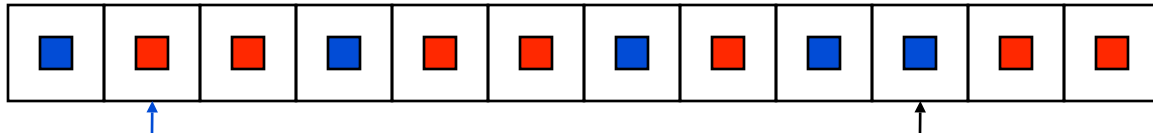
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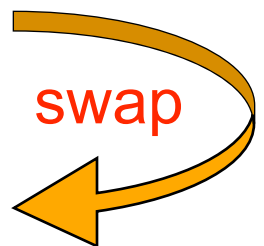
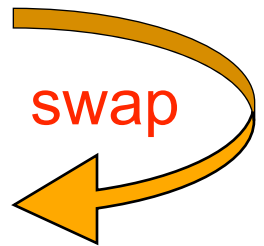
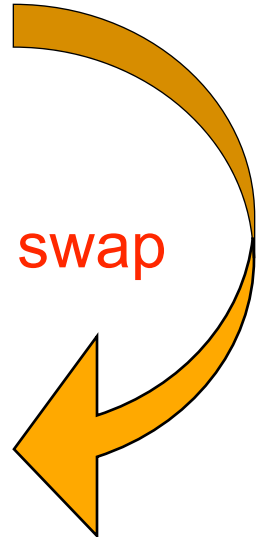
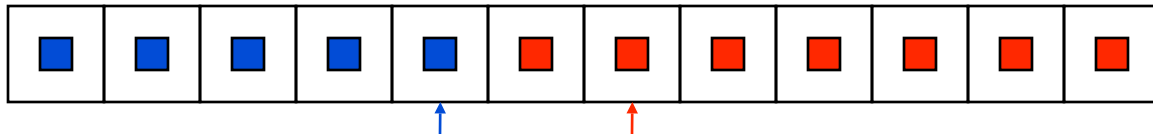
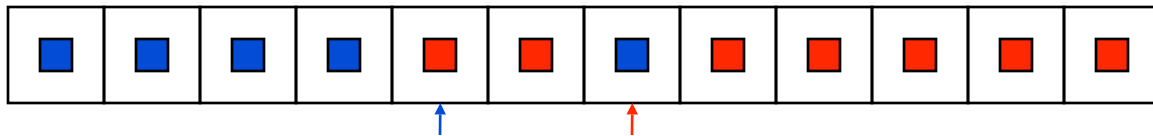
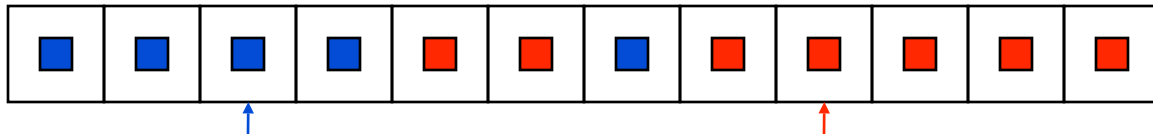
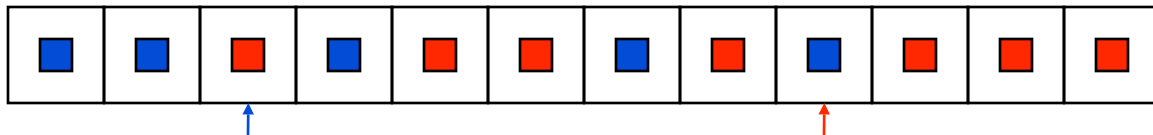
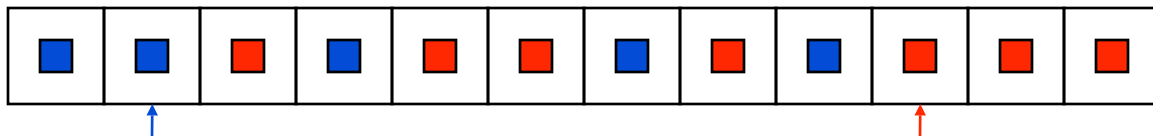


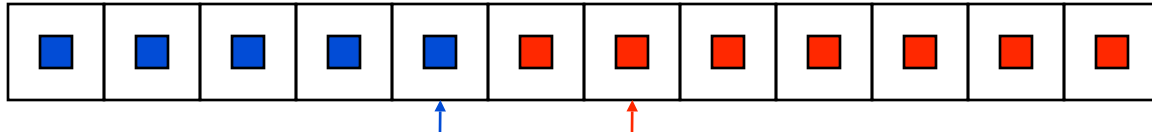
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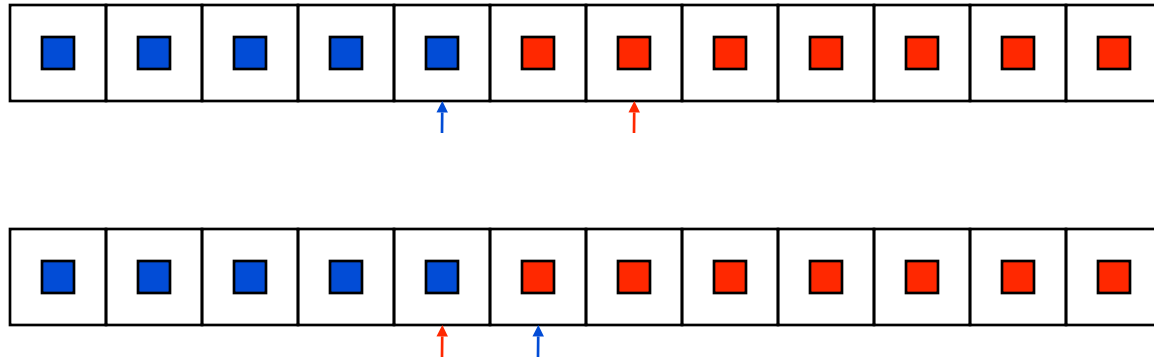


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- Once indices cross, partitioning is done
- If you replace blue with $\leq p$ and red with $\geq p$, this is exactly what we need for QuickSort partitioning
- Notice that after partitioning, array is partially sorted
- Recursive calls on partitioned subarrays will sort subarrays
- No need to copy/move arrays, since we partitioned in place

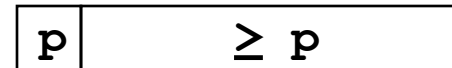


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QuickSort Analysis

- Runtime analysis (worst-case)

- Partition can work badly, producing this:



- Runtime recurrence

- ♦ $T(n) = T(n-1) + n$

- This can be solved to show worst-case $T(n)$ is $O(n^2)$

- Runtime analysis (expected-case)

- More complex recurrence

- Can solve to show *expected* $T(n)$ is $O(n \log n)$

- Improve constant factor by avoiding **QuickSort** on small sets

- Switch to **InsertionSort** (for example) for sets of size, say, ≤ 9

- Definition of *small* depends on language, machine, etc.

Sorting Algorithm Summary

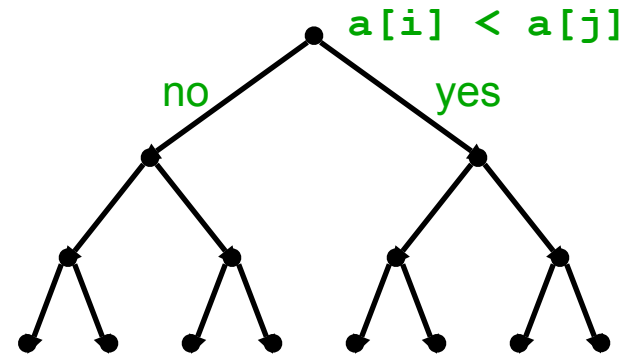
- The ones we have discussed
 - InsertionSort
 - SelectionSort
 - MergeSort
 - QuickSort
- Other sorting algorithms
 - HeapSort (will revisit this)
 - ShellSort (in text)
 - BubbleSort (nice name)
 - RadixSort
 - BinSort
 - CountingSort
- Why so many? Do computer scientists have some kind of sorting fetish or what?
 - Stable sorts: **Ins, Sel, Mer**
 - Worst-case $O(n \log n)$: **Mer, Hea**
 - Expected $O(n \log n)$: **Mer, Hea, Qui**
 - Best for nearly-sorted sets: **Ins**
 - No extra space needed: **Ins, Sel, Hea**
 - Fastest in practice: **Qui**
 - Least data movement: **Sel**

Lower Bound for Comparison Sorting

- Goal: Determine the minimum time *required* to sort n items
- Note: we want *worst-case*, not *best-case* time
 - Best-case doesn't tell us much; for example, we know Insertion Sort takes $O(n)$ time on already-sorted input
 - Want to know the *worst-case time* for the *best possible* algorithm
- But how can we prove anything about the *best possible* algorithm?
 - We want to find characteristics that are common to *all* sorting algorithms
 - Let's limit attention to *comparison-based algorithms* and try to count number of comparisons

Comparison Trees

- Comparison-based algorithms make decisions based on comparison of data elements
- This gives a *comparison tree*
- If the algorithm fails to terminate for some input, then the comparison tree is infinite
- The height of the comparison tree represents the *worst-case number of comparisons* for that algorithm
- Can show that *any* correct comparison-based algorithm must make at least $n \log n$ comparisons in the worst case



Lower Bound for Comparison Sorting

- Say we have a correct comparison-based algorithm
- Suppose we want to sort the elements in an array $B[]$
- Assume the elements of $B[]$ are distinct
- Any permutation of the elements is initially possible
- When done, $B[]$ is sorted
- But the algorithm could not have taken the same path in the comparison tree on different input permutations

Lower Bound for Comparison Sorting

- How many input permutations are possible? $n! \sim 2^{n \log n}$
- For a comparison-based sorting algorithm to be correct, it must have at least that many leaves in its comparison tree
- to have at least $n! \sim 2^{n \log n}$ leaves, it must have height at least $n \log n$ (since it is only binary branching, the number of nodes at most doubles at every depth)
- therefore its longest path must be of length at least $n \log n$, and that is its worst-case running time

java.lang.Comparable<T> Interface

- `public int compareTo(T x) ;`
 - Returns a negative, zero, or positive value
 - ◆ negative if `this` is before `x`
 - ◆ 0 if `this.equals(x)`
 - ◆ positive if `this` is after `x`
- Many classes implement `Comparable`
 - `String`, `Double`, `Integer`, `Character`, `Date`,...
 - If a class implements `Comparable`, then its `compareTo` method is considered to define that class's *natural ordering*
- Comparison-based sorting methods should work with `Comparable` for maximum generality