



CSE373: Data Structures and Algorithms

## Comparison Sorting

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Winter 2016

This lecture material represents the work of multiple instructors at the University of Washington. Thank you to all who have contributed!

## Introduction to Sorting

- Stacks, queues, priority queues, and dictionaries all focused on providing one element at a time
- But often we know we want "all the things" in some order
  - Humans can sort, but computers can sort fast
  - Very common to need data sorted somehow
    - Alphabetical list of people
    - List of countries ordered by population
    - Search engine results by relevance
    - ...
- Algorithms have different asymptotic and constant-factor trade-offs
  - No single "best" sort for all scenarios
  - Knowing one way to sort just isn't enough

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## More Reasons to Sort

General technique in computing:

*Preprocess data to make subsequent operations faster*

Example: Sort the data so that you can

- Find the  $k^{\text{th}}$  largest in constant time for any  $k$
- Perform binary search to find elements in logarithmic time

Whether the performance of the preprocessing matters depends on

- How often the data will change (and how much it will change)
- How much data there is

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## Why Study Sorting in this Class?

- You might never need to reimplement a sorting algorithm yourself
  - Standard libraries will generally implement one or more (Java implements 2)
- You will almost certainly use sorting algorithms
  - Important to understand relative merits and expected performance
- Excellent set of algorithms for practicing analysis and comparing design techniques
  - Classic part of a data structures class, so you'll be expected to know it

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## The main problem, stated carefully

For now, assume we have  $n$  comparable elements in an array and we want to rearrange them to be in increasing order

Input:

- An array  $A$  of data records
- A key value in each data record
- A comparison function (consistent and total)

Effect:

- Reorganize the elements of  $A$  such that for any  $i$  and  $j$ , if  $i < j$  then  $A[i] \leq A[j]$
- (Also,  $A$  must have exactly the same data it started with)
- Could also sort in reverse order, of course

An algorithm doing this is a [comparison sort](#)

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## Variations on the Basic Problem

1. Maybe elements are in a linked list (could convert to array and back in linear time, but some algorithms needn't do so)
2. Maybe ties need to be resolved by "original array position"
  - Sorts that do this naturally are called [stable sorts](#)
  - Others could tag each item with its original position and adjust comparisons accordingly (non-trivial constant factors)
3. Maybe we must not use more than  $O(1)$  "auxiliary space"
  - Sorts meeting this requirement are called [in-place sorts](#)
4. Maybe we can do more with elements than just compare
  - Sometimes leads to faster algorithms
5. Maybe we have too much data to fit in memory
  - Use an ["external sorting"](#) algorithm

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### Sorting: The Big Picture

Surprising amount of neat stuff to say about sorting:

<b>Simple algorithms:</b> $O(n^2)$	<b>Fancier algorithms:</b> $O(n \log n)$	<b>Comparison lower bound:</b> $\Omega(n \log n)$	<b>Specialized algorithms:</b> $O(n)$	<b>Handling huge data sets</b>
Insertion sort Selection sort Shell sort ...	Heap sort Merge sort Quick sort ...		Bucket sort Radix sort	External sorting

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### Insertion Sort

- Idea: At step  $k$ , put the  $k^{\text{th}}$  element in the correct position among the first  $k$  elements
- Alternate way of saying this:
  - Sort first two elements
  - Now insert 3<sup>rd</sup> element in order
  - Now insert 4<sup>th</sup> element in order
  - ...
- "Loop invariant": when loop index is  $i$ , first  $i$  elements are sorted
- Let's see a visualization (<http://www.cs.uafca.edu/~galles/visualization/ComparisonSort.html>)
- Time?
  - Best-case \_\_\_\_\_ Worst-case \_\_\_\_\_ "Average" case \_\_\_\_\_

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- Time?
  - Best-case  $O(n)$  Worst-case  $O(n^2)$  "Average" case  $O(n^2)$
  - start sorted start reverse sorted (see text)

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### Selection sort

- Idea: At step  $k$ , find the smallest element among the not-yet-sorted elements and put it at position  $k$
- Alternate way of saying this:
  - Find smallest element, put it 1<sup>st</sup>
  - Find next smallest element, put it 2<sup>nd</sup>
  - Find next smallest element, put it 3<sup>rd</sup> ...
- "Loop invariant": when loop index is  $i$ , first  $i$  elements are the  $i$  smallest elements in sorted order
- Let's see a visualization (<http://www.cs.uafca.edu/~galles/visualization/ComparisonSort.html>)
- Time?
  - Best-case \_\_\_\_\_ Worst-case \_\_\_\_\_ "Average" case \_\_\_\_\_

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- Time?
  - Best-case  $O(n^2)$  Worst-case  $O(n^2)$  "Average" case  $O(n^2)$
  - Always  $T(1) = 1$  and  $T(n) = n + T(n-1)$

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### Insertion Sort vs. Selection Sort

- Different algorithms
- Solve the same problem
- Have the same worst-case and average-case asymptotic complexity
  - Insertion-sort has better best-case complexity; preferable when input is "mostly sorted"
- Other algorithms are more efficient for *large arrays that are not already almost sorted*
  - Insertion sort may do well on small arrays

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### Aside: We Will Not Cover Bubble Sort

- It is not, in my opinion, what a “normal person” would think of
- It doesn't have good asymptotic complexity:  $O(n^2)$
- It's not particularly efficient with respect to constant factors

Basically, almost everything it is good at some other algorithm is at least as good at

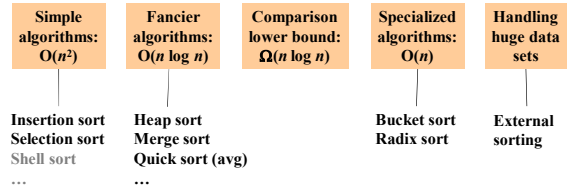
- Perhaps people teach it just because someone taught it to them?

Fun, short, optional read:

*Bubble Sort: An Archaeological Algorithmic Analysis*, Owen Astrachan, SIGCSE 2003, <http://www.cs.duke.edu/~ola/bubble/bubble.pdf>

### The Big Picture

Surprising amount of juicy computer science: 2-3 lectures...



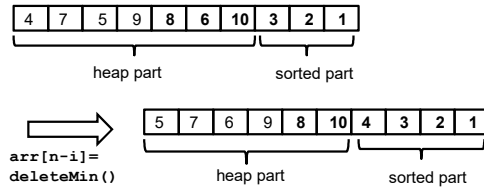
### Heap sort

- Sorting with a heap is easy:
  - `insert` each `arr[i]`, or better yet use `buildHeap`
  - `for i in range(len(arr)):`  
`arr[i] = deleteMin()`
- Worst-case running time:  $O(n \log n)$
- We have the array-to-sort and the heap
  - So this is not an in-place sort
  - There's a trick to make it in-place...

### In-place heap sort

But this reverse sorts – how would you fix that?

- Treat the initial array as a heap (via `buildHeap`)
- When you delete the  $i^{\text{th}}$  element, put it at `arr[n-i]`
  - That array location isn't needed for the heap anymore!



### “AVL sort”

- We can also use a balanced tree to:
  - `insert` each element: total time  $O(n \log n)$
  - Repeatedly `deleteMin`: total time  $O(n \log n)$ 
    - Better: in-order traversal  $O(n)$ , but still  $O(n \log n)$  overall
- But this cannot be made in-place and has worse constant factors than heap sort
  - both are  $O(n \log n)$  in worst, best, and average case
  - neither parallelizes well
  - heap sort is better

### “Hash sort”???

- Don't even think about trying to sort with a hash table!
- Finding min item in a hashtable is  $O(n)$ , so this would be a slower, more complicated selection sort
- And we've already seen that selection sort is pretty bad!

### Divide and conquer

Very important technique in algorithm design

1. Divide problem into smaller parts
2. Independently solve the simpler parts
  - Think recursion
  - Or potential parallelism
3. Combine solution of parts to produce overall solution

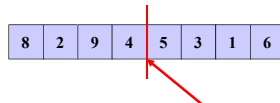
(This technique has a *long* history.)

### Divide-and-Conquer Sorting

Two great sorting methods are fundamentally divide-and-conquer

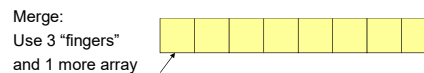
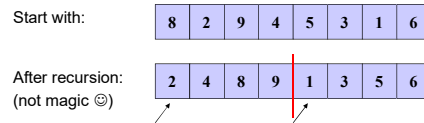
1. Mergesort: Sort the left half of the elements (recursively)  
Sort the right half of the elements (recursively)  
Merge the two sorted halves into a sorted whole
2. Quicksort: Pick a "pivot" element  
Divide elements into less-than pivot  
and greater-than pivot  
Sort the two divisions (recursively on each)  
Answer is sorted-less-than then pivot then sorted-greater-than

### Merge sort



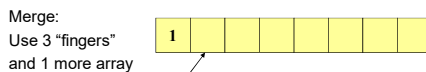
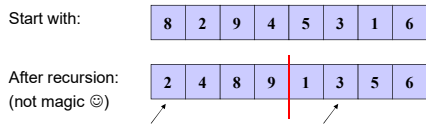
- To sort array from position  $l_0$  to position  $h_1$ :
  - If range is 1 element long, it is already sorted! (Base case)
  - Else:
    - Sort from  $l_0$  to  $(h_1+l_0)/2$
    - Sort from  $(h_1+l_0)/2$  to  $h_1$
    - Merge the two halves together
- Merging takes two sorted parts and sorts everything
  - $O(n)$  but requires auxiliary space...

### Example, focus on merging



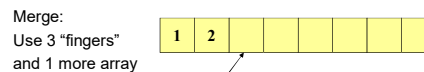
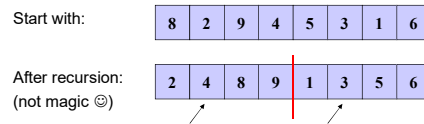
(After merge, copy back to original array)

### Example, focus on merging



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**Example, focus on merging**

Start with: 

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

After recursion: (not magic ☺) 

2	4	8	9	1	3	5	6
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Merge: Use 3 "fingers" and 1 more array

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After recursion: (not magic ☺) 


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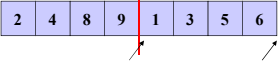
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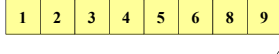
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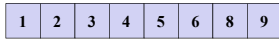
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### Example, focus on merging

Start with: 

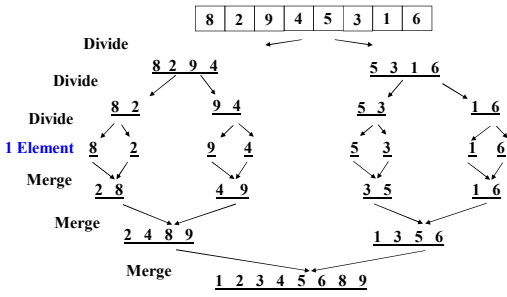
After recursion: (not magic ☺) 

Merge: Use 3 "fingers" and 1 more array 

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### Example, Showing Recursion



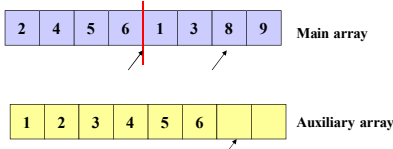
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### Merge sort visualization

- <http://www.cs.usfca.edu/~galles/visualization/ComparisonSort.html>

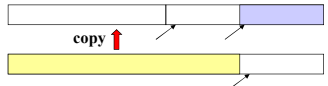
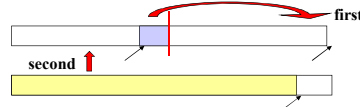
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### Some details: saving a little time

- What if the final steps of our merge looked like this:
 
- Wasteful to copy to the auxiliary array just to copy back...

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### Some details: saving a little time

- If left-side finishes first, just stop the merge and copy back:
 
- If right-side finishes first, copy dregs into right then copy back:
 

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### Some details: Saving Space and Copying

Simplest / Worst:  
Use a new auxiliary array of size  $(hi - lo)$  for every merge

Better:  
Use a new auxiliary array of size  $n$  for every merging stage

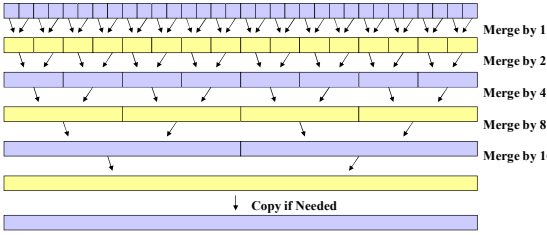
Better:  
Reuse same auxiliary array of size  $n$  for every merging stage

Best (but a little tricky):  
Don't copy back – at 2<sup>nd</sup>, 4<sup>th</sup>, 6<sup>th</sup>, ... merging stages, use the original array as the auxiliary array and vice-versa  
– Need one copy at end if number of stages is odd

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### Swapping Original / Auxiliary Array ("best")

- First recurse down to lists of size 1
- As we return from the recursion, swap between arrays



(Arguably easier to code up without recursion at all)

### Linked lists and big data

We defined sorting over an array, but sometimes you want to sort linked lists

One approach:

- Convert to array:  $O(n)$
- Sort:  $O(n \log n)$
- Convert back to list:  $O(n)$

Or merge sort works very nicely on linked lists directly

- Heapsort and quicksort do not
- Insertion sort and selection sort do but they're slower

Merge sort is also the sort of choice for external sorting

- Linear merges minimize disk accesses
- And can leverage multiple disks to get streaming accesses

### Analysis

Having defined an algorithm and argued it is correct, we should analyze its running time and space:

To sort  $n$  elements, we:

- Return immediately if  $n=1$
- Else do 2 subproblems of size  $n/2$  and then an  $O(n)$  merge

Recurrence relation:

$$T(1) = c_1$$

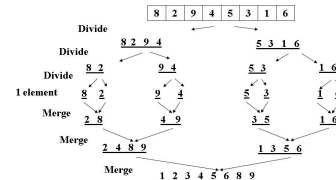
$$T(n) = 2T(n/2) + c_2n$$

### Analysis intuitively

This recurrence is common you just "know" it's  $O(n \log n)$

Merge sort is relatively easy to intuit (best, worst, and average):

- The recursion "tree" will have  $\log n$  height
- At each level we do a *total* amount of merging equal to  $n$



### Analysis more formally (One of the recurrence classics)

For simplicity let constants be 1 (no effect on asymptotic answer)

$$T(1) = 1$$

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

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

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

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

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

$$\dots$$

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

So total is  $2^k T(n/2^k) + kn$  where  $n/2^k = 1$ , i.e.,  $\log n = k$

That is,  $2^{\log n} T(1) + n \log n$

$$= n + n \log n$$

$$= O(n \log n)$$

### Next lecture

- Quick sort ☺