

---

# Quicksort

## Lower Bounds for Comparison Sorts

CS 4102, Algorithms  
Prof. Floryan and Prof. Horton  
Spring 2021

# Quicksort and Partition

Readings:  
CLRS Chapter 7 (not 7.4.2)

# Quicksort: Introduction

---

- ▶ Developed by C.A.R. (Tony) Hoare (a Turing Award winner)  
[http://www.wikipedia.org/wiki/C.\\_A.\\_R.\\_Hoare](http://www.wikipedia.org/wiki/C._A._R._Hoare)
  - ▶ Published in 1962
- ▶ Classic divide and conquer, but...
  - ▶ Mergesort does no comparisons to divide, but a lot to combine results (i.e. the merge) at each step
  - ▶ Quicksort does a lot of work to divide, but has nothing to do after the recursive calls. No work to combine.
    - ▶ If we're using arrays. Linked lists? Interesting to think about this!
- ▶ Dividing done with algorithm often called ***partition***
  - ▶ Sometimes called *split*. Several variations.

# Quicksort's Strategy

---

- ▶ Called on subsection of array from *first* to *last*
  - ▶ Like mergesort
- ▶ First, choose some element in the array to be the ***pivot*** element
  - ▶ Any element! Doesn't matter for correctness.
  - ▶ Often the first item. For us, the last. Or, we often move some element into the last position (to get better efficiency)
- ▶ Second, call ***partition***, which does two things:
  - ▶ Puts the pivot in its proper place, i.e. where it will be in the correctly sorted sequence
  - ▶ All elements below the pivot are less-than the pivot, and all elements above the pivot are greater-than
- ▶ Third, use quicksort recursively on both sub-lists

# Quicksort is Divide and Conquer

---

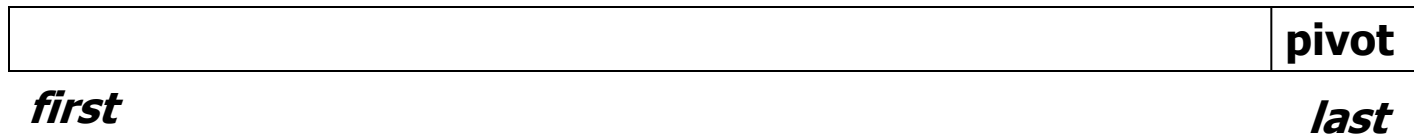
- ▶ **Divide:** select **pivot** element  $p$ , **Partition**( $p$ )
- ▶ **Conquer:** recursively sort left and right sublists
- ▶ **Combine:** Nothing!

Contrast to mergesort,  
where divide is simple and combine is work

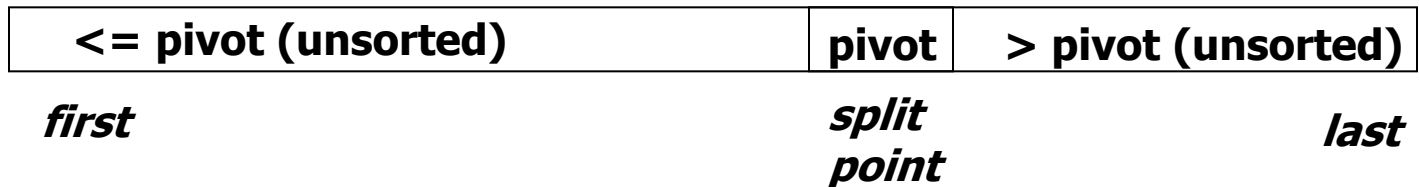
# Quicksort's Strategy (a picture)

---

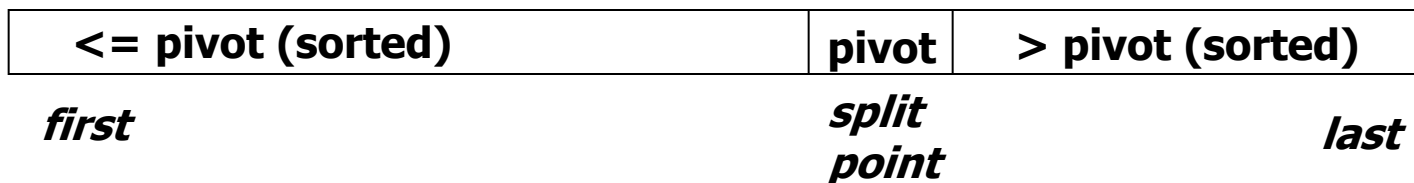
- ▶ Use last element as pivot (or pick one and move it there)



- ▶ After call to partition...



- ▶ Now sort two parts recursively and we're done!



- ▶ Note that splitPoint may be anywhere in *first..last*
- ▶ Note our assumption that all keys are distinct

# Quicksort Code

---

Input Parameters: *list, first, last*

Output Parameters: *list*

```
def quicksort(list, first, last):  
    if first < last:  
        q = partition(list, first, last)  
        quicksort(list, first, q-1)  
        quicksort(list, q+1, last)  
    return
```

# Partition Does the Dirty Work

---

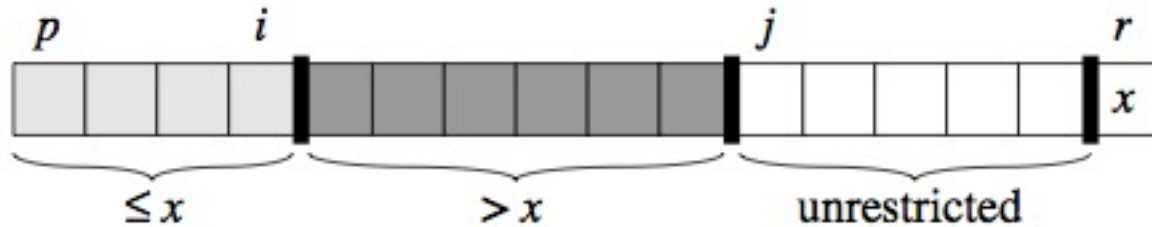
- ▶ Partition rearranges elements
  - ▶ How? How many comparisons? How many swaps?
- ▶ How? Two well-known algorithms
  - ▶ In this chapter of CLRS, Lomuto's algorithm
  - ▶ In the exercises, the original: Hoare's algorithm.  
(Page 185. Look at on your own.)
  - ▶ Important:
    - ▶ Both are in-place!
    - ▶ Both are linear.



# Strategy for Lomuto's Partition

- Invariant: At any point:
  - $i$  indexes the right-most element  $\leq pivot$
  - $j-1$  indexes the right-most element  $> pivot$

**Note:** book uses  $p$  and  $r$  for what we called *first* and *last* in earlier slides



- ▶ **Strategy:**
  - ▶ Look at next item  $a[j]$
  - ▶ If that item  $> pivot$ , all is well!
  - ▶ If that item  $< pivot$ , increment  $i$  and then swap items at positions  $i$  and  $j$
  - ▶ When done, swap pivot with item at position  $i+1$
- ▶ **Number of comparisons:**  $n-1$

# Efficiency of Quicksort

---

- ▶ Partition divides into two sub-lists, perhaps unequal size
  - ▶ Depends on value of pivot element
- ▶ Recurrence for Quicksort
$$T(n) = \text{partition-cost} + T(\text{size of 1st section}) + T(\text{size of 2nd section})$$
- ▶ If divides equally,  $T(n) = 2 T(n/2) + n-1$ 
  - ▶ Just like mergesort
  - ▶ Solve by substitution or master theorem
$$T(n) \in \Theta(n \lg n)$$
- ▶ This is the best-case. But...

# Worst Case of Quicksort

---

- ▶ What if divides in most unequal fashion possible?
  - ▶ One subsection has size 0, other has size  $n-1$
  - ▶  $T(n) = T(0) + T(n-1) + n-1$
  - ▶ What if this happens every time we call partition recursively?

$$W(n) = \sum_{k=2}^n (k-1) \in \Theta(n^2)$$

- ▶ Uh oh. Same as insertion sort.
  - ▶ “Sorry Prof. Hoare – we have to take back that Turing Award now!”

# Quicksort's Average Case

---

- ▶ Good if it divides equally, bad if most unequal.
  - ▶ Remember: when subproblems size 0 and  $n-1$
  - ▶ Can worst-case happen?  
Sure! Many cases. One is when elements already sorted. Last element is max, pivot around that. Next pivot is 2<sup>nd</sup> max...
- ▶ What's the average?
  - ▶ Much closer to the best case
  - ▶ A bad-split then a good-split is closer to best-case (pp. 176-178)
  - ▶ To prove  $A(n)$ , fun with recurrences!
  - ▶ The result: If all permutations are equal, then  
 $A(n) \cong 1.386 n \lg n$  (for large  $n$ )
- ▶ So very fast on average.
- ▶ And, we can take simple steps to avoid the worst case!

# Avoiding Quicksort's Worst Case

---

- ▶ Make sure we don't pivot around max or min
  - ▶ Find a better choice and swap it with last element
  - ▶ Then partition as before
- ▶ Recall we get best case if divides equally
  - ▶ Could find median. But this costs  $\Theta(n)$ . Instead...
  - ▶ Choose a **random element** between first and last and swap it with the last element
  - ▶ Or, estimate the median by using the “median-of-three” method
    - ▶ Pick 3 elements (say, first, middle and last)
    - ▶ Choose median of these and swap with last. (Cost?)
    - ▶ If sorted, then this chooses real median. Best case!

# Tuning Quicksort's Performance

---

- ▶ In practice quicksort runs fast
  - ▶  $A(n)$  is log-linear, and the “constants” are smaller than mergesort and heapsort
  - ▶ Often used in software libraries
  - ▶ So worth tuning it to squeeze the most out of it
  - ▶ Always do something to avoid worst-case
- ▶ Sort small sub-lists with (say) insertion sort
  - ▶ For small inputs, insertion sort is fine
    - ▶ No recursion, function calls
  - ▶ Variation: don't sort small sections at all.  
After quicksort is done, sort entire array with **insertion sort**
    - ▶ It's efficient on almost-sorted arrays!

# Quicksort's Space Complexity

---

- ▶ Looks like it's in-place, but there's a *recursion stack*
  - ▶ Depends on your definition: some people define *in-place* to **not** include stack space used by recursion
    - ▶ E.g. our CLRS algorithms textbook
    - ▶ Other books and people do "count" this
  - ▶ How much goes on the stack?
    - ▶ If most uneven splits, then  $\Theta(n)$ .
    - ▶ If splits evenly every time, then  $\Theta(\lg n)$ .
- ▶ Ways to reduce stack-space used due to recursion
  - ▶ Various books cover the details (not ours, though)
  - ▶ First, remove 2nd recursive call (tail-recursion)
  - ▶ Second, always do recursive call on smaller section

# Summary: Quicksort

---

- ▶ Divide and conquer where divide does the heavy-lifting
- ▶ In worst-case, efficiency is  $\Theta(n^2)$ 
  - ▶ But it's practical to avoid the worst-case
- ▶ On average, efficiency is  $\Theta(n \lg n)$
- ▶ Better space-complexity than mergesort.
- ▶ In practice, runs fast and widely used
  - ▶ Many ways to tune its performance
- ▶ Various strategies for Partition
  - ▶ Some work better if duplicate keys
- ▶ More details? See Sedgewick's algorithms textbook
  - ▶ He's the expert! PhD on this under Donald Knuth



# Lower Bounds Proof for Comparison Sorts

Readings: CLRS Section 8.1

## Mental Stretch

---

Show  $\log(n!) \in \Theta(n \log n)$

Hint: show  $n! \leq n^n$

Hint 2: show  $n! \geq \left(\frac{n}{2}\right)^{\frac{n}{2}}$

$$\log n! \in O(n \log n)$$

$$\begin{array}{ccccccc} n! & = & n & \cdot & (n-1) & \cdot & (n-2) & \cdot & \dots & \cdot & 2 & \cdot & 1 \\ & & \parallel & & \wedge & & \wedge & & & & \wedge & & \wedge \\ n^n & = & n & \cdot & n & \cdot & n & \cdot & \dots & \cdot & n & \cdot & n \end{array}$$

---

$$\begin{aligned} n! &\leq n^n \\ \Rightarrow \log(n!) &\leq \log(n^n) \\ \Rightarrow \log(n!) &\leq n \log n \\ \Rightarrow \log(n!) &\in O(n \log n) \end{aligned}$$



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

$$n! = n \cdot (n-1) \cdot (n-2) \cdot \dots \cdot \frac{n}{2} \cdot \left(\frac{n}{2}-1\right) \cdot \dots \cdot 2 \cdot 1$$

$$\begin{array}{cccccccc} & \vee & \vee & \vee & \parallel & \vee & & \vee \parallel \\ \left(\frac{n}{2}\right)^{\frac{n}{2}} & = & \frac{n}{2} \cdot & \frac{n}{2} & \cdot & \frac{n}{2} & \cdot \dots \cdot \frac{n}{2} \cdot & 1 & \cdot \dots \cdot 1 \cdot 1 \end{array}$$


---

$$n! \geq \left(\frac{n}{2}\right)^{\frac{n}{2}}$$

$$\Rightarrow \log(n!) \geq \log\left(\left(\frac{n}{2}\right)^{\frac{n}{2}}\right)$$

$$\Rightarrow \log(n!) \geq \frac{n}{2} \log \frac{n}{2}$$

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



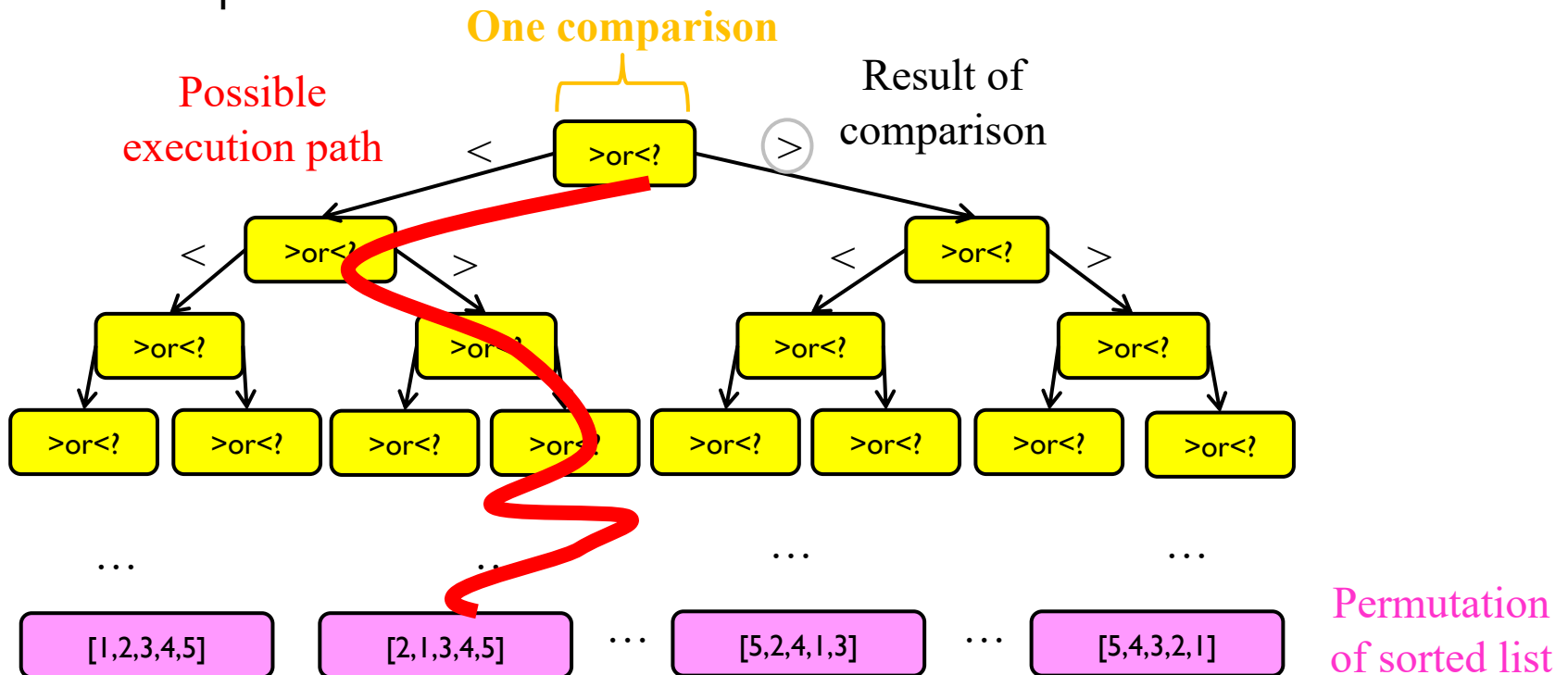
# Worst Case Lower Bounds

---

- ▶ Prove that there is no algorithm which can sort faster than  $O(n \log n)$
- ▶ Non-existence proof!
  - ▶ Seems like maybe it would be very hard to do... (?)

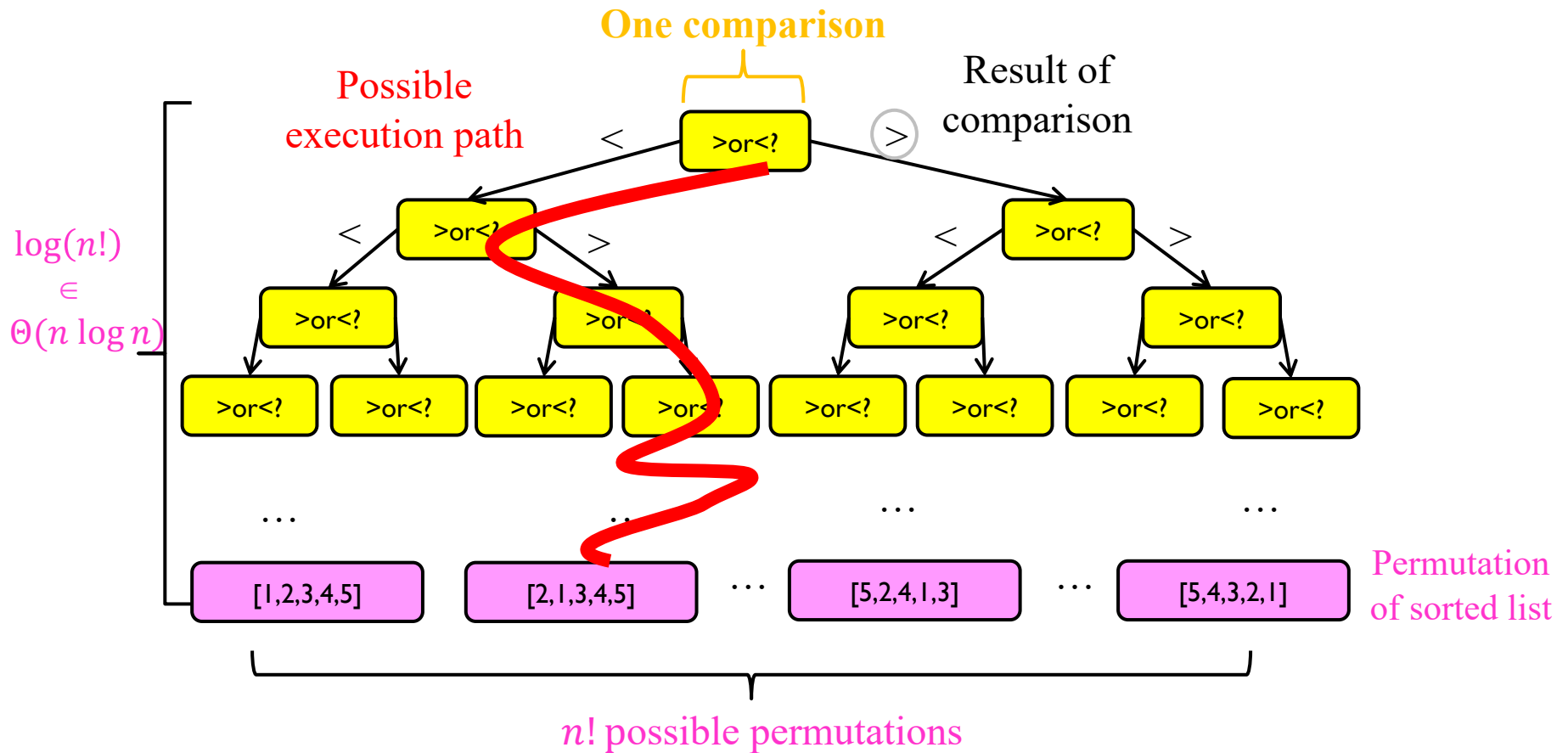
# Strategy: Decision Tree

- ▶ Sorting algorithms use comparisons to determine the order of input elements
- ▶ Conceptually possible to draw a tree to illustrate all possible execution paths



# Strategy: Decision Tree

- ▶ Worst case run time is the longest execution path
- ▶ i.e., “height” of the decision tree



# Lower Bound for Worst Case

---

- ▶ Binary tree property: At level  $d$  in a binary tree, there are at most  $2^d$  nodes (where level of root is 0)
- ▶ Also, let's say a tree's height is number of levels minus one
  - ▶ Height of our decision tree is the  $W(n)$  number of comparisons
- ▶ Theorem 8.1 (p. 193):
  - ▶ Let  $L$  be the number of leaves in a binary tree and let  $h$  be its height. (Book uses lower-case  $l$ , not  $L$  like we do here.)
  - ▶ Then  $L \leq 2^h$ . (Number of leaves is no more than  $2^h$ .)
  - ▶ Therefore  $h \geq \lceil \lg L \rceil$  (Height is not less than...)
  - ▶ For a correct sorting algorithm,  $L \geq n!$
  - ▶ Therefore
$$h \geq \lceil \lg L \rceil \geq \lceil \lg n! \rceil$$
- ▶ Thus, for any algorithm that sorts by comparison of keys  
 $W(n)$  is at least  $\lceil \lg n! \rceil$



# Formula for the Lower Bound

---

- ▶ Earlier we showed this was  $\Theta(n \lg n)$
- ▶ Or, we can we lose that factorial in other ways
  - ▶ Stirling's formula:  $(n/e)^n \sqrt{2\pi n}$ 
    - ▶ Take the log of this approximation of  $n!$  and you'll see that it's  $\Theta(n \lg n)$
  - ▶ Better to re-write, use integrals, and...
    - ▶ See a textbook for details (but not ours)
- ▶ If you were to do all this, you'd see:

$$W(n) \geq \lceil \lg n! \rceil \geq \lceil n \lg n - 1.443n \rceil$$

which is of course  $\Theta(n \lg n)$

- ▶ FYI Mergesort is very close to optimal
  - ▶ But not for all values of  $n$

# Summary

---

- ▶ Our lower-bound proof shows any algorithm must be  $\Omega(n \lg n)$  in the worst-case if it works by comparing keys
  - ▶ Algorithms that only do key-comparisons can sort any data type
  - ▶ Algorithms that can calculate on their keys can do better
    - ▶ E.g. counting sort and radix sort for numbers (Ch. 8 of CLRS)
  - ▶ In the same way that binary search is optimal, but hashing can be faster
- ▶ Mergesort and Quicksort are in this order-class
  - ▶ Mergesort is very close to the L.B. (but not in-place)
  - ▶ But quicksort will run faster generally
    - ▶ Why? Constants and lower-order terms are smaller. In other words, the overhead per comparison is less.
  - ▶ But Quicksort really could be  $\Theta(n^2)$  at its worst