CSC 148: Introduction to Computer Science Week 11

Efficiency considerations
Sorting efficiency



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sorting

How does the time to sort a list with n elements vary with n?

- 108 sorts:
 - bubble sort -> n²
 - selection sort -> n²
 - insertion sort -> n²



sorting

- How does the time to sort a list with n elements vary with n?
- 108 sorts:
 - bubble sort -> n²
 - selection sort -> n²
 - insertion sort -> n²
- Some other sort?
 - Merge sort?
 - Quicksort?
 - Radix sorts?
 - Apropos of nothing: Andrew's favorite for sorting exams



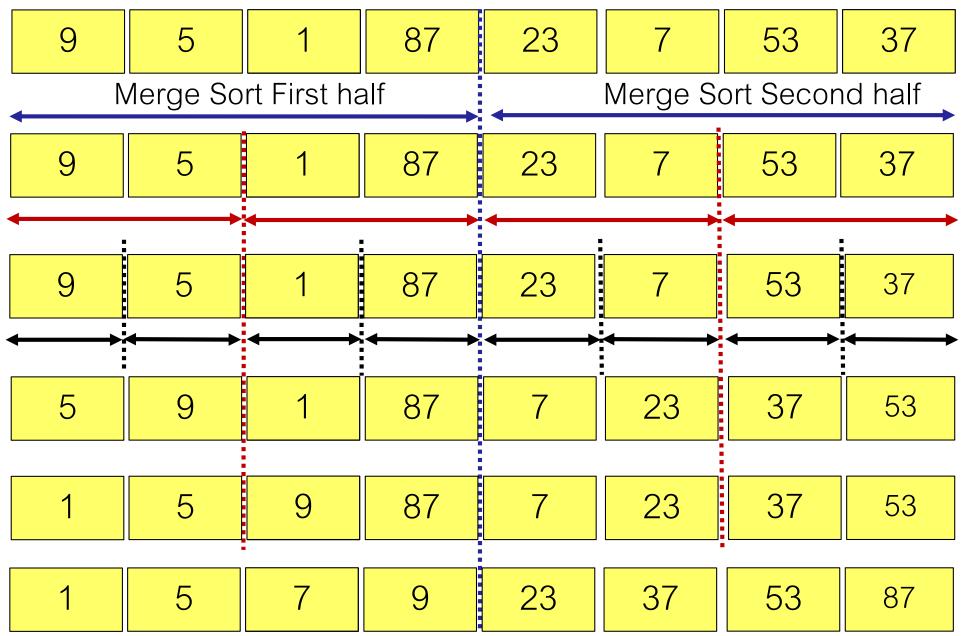
Merge Sort

Recursive algorithm based on the idea of "divide and conquer"

- Split a list in two halves repeatedly
- Halves with 0 or 1 elements are guaranteed sorted
- Merge the two halves "on the way back"
 - All the work is in merging!

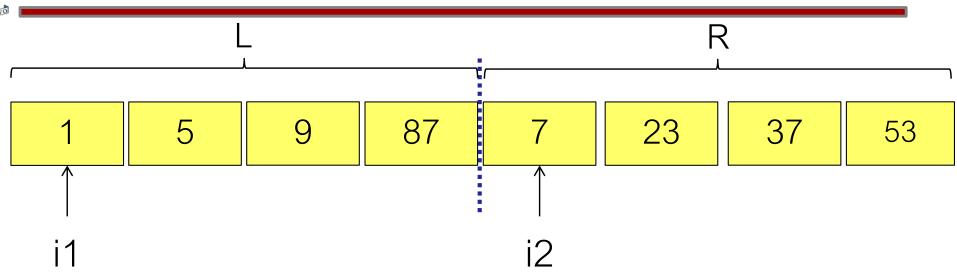


Merge Sort: split all the way, then merge

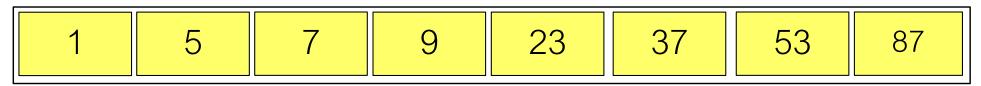




Merge step: merge(L, R)



sorted list (different than L or R)



The halves might not be perfectly equal though...



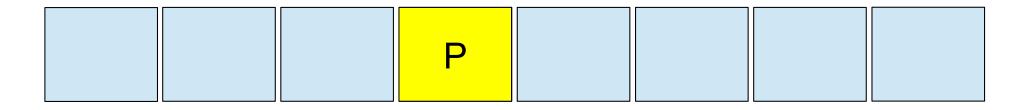
Another "divide and conquer" algorithm – but doesn't necessarily split into *half*.

- Split a list ("partitioning") into the part smaller than some value (called pivot) and the part not smaller than that value
- Sort these two parts
- Recombine the list

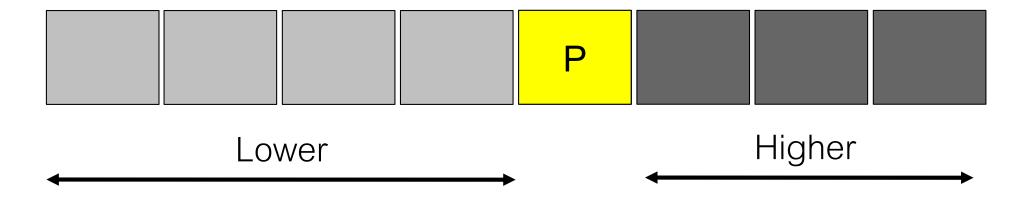


Partitioning

Begin with the unsorted list and select a pivot P at random



 Split list such that all elements to the left are lower than P and all to the right are higher

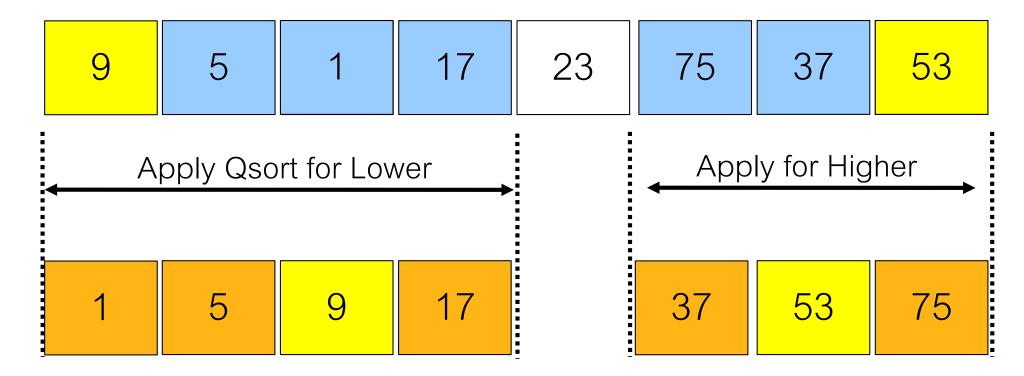


Several ways to do the partition step ...



Quicksort Recursion

- Recurse: repeat the same idea for the two partitions
- Pick pivot, process such that all lower than it are on the left, all higher on the right





Worksheet!

- How do we analyse the running time of recursive algorithms in general? (Not just for trees.)
- Two key parts:
 - how long do the non-recursive parts take?
 - what is the structure of the recursive calls?



Merge sort

 idea: break a list up (partition) into two halves, merge sort each half, then recombine (merge) the halves

```
def mergesort(lst: list) -> list:
  """Return a sorted list with the same elements as <|st>.
  This is a *non-mutating* version of mergesort; it does not mutate the
  input list.
                                          Lists of length < 2 are
  111111
  if len(lst) < 2:
                                              already sorted
      return lst[:]
  else:
                                                                   First half
      mid = len(lst) // 2
       left_sorted = mergesort(lst[:mid])
       right_sorted = mergesort(lst[mid:])
                                                                 Second half
       return _merge(left_sorted, right_sorted)
                           Merge the two sorted halves,
                             "on the way back". How?
```



Counting Merge Sort

- Assume a list of size n
- Merge operation takes linear time ... why?
- The "divide" step also takes linear time (approx n steps) ... why?
- What about the cost of the two recursive calls?



Counting Merge Sort: *n* = 8

```
ms([4, 2, 6, 8, 1, 3, 5, 7])
 prop. to log n
      splits
                       merge( ms([4, 2, 6, 8]) , ms([1, 3, 5, 7]) )
               merge( merge(ms([4,2]), ms([6,8])) , merge(ms([1,3]), ms([5,7])
merge(merge(merge(ms([4]),ms([2])),merge(ms([6]),ms([8])), merge(merge(ms([1]),ms([3])),merge(ms([5]),ms([7])
       merge( merge([4],[2]),merge([6],[8])), merge(merge([1],[3]),merge([5],[7])) ) ),
                           merge( merge([2,4],[6,8]), merge([1,3],[5,7])
                                     merge([2,4,6,8], [1,3,5,7])
                                                                               => n \times log n
 log n merge with
                                        [1, 2, 3, 4, 5, 6, 7, 8]
prop. to n copies
```

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Merge Sort

```
def mergesort(lst):
    if len(lst) < 2:
        return lst[:]
    else:
        mid = len(lst) // 2
        left = lst[:mid]
        right = lst[mid:]
        left sorted = mergesort(left)
        right sorted = mergesort(right)
        return merge (left sorted, right sorted)
```

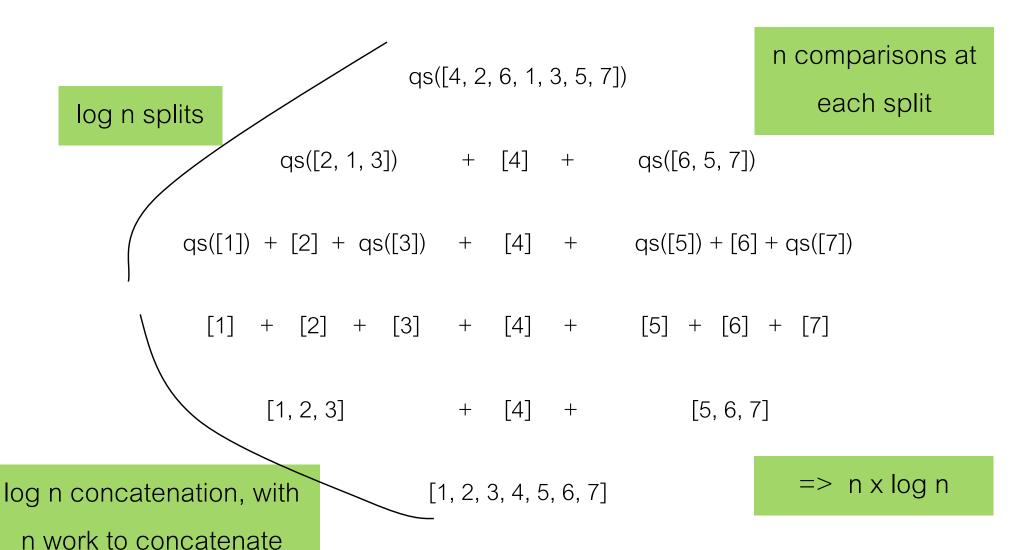


 idea: break a list up (partition) into the part smaller than some value (pivot) and not smaller than that value, sort these parts, then recombine the list:

```
def quicksort(lst: list) -> list:
       Return a sorted list with the same elements as <|st>.
       This is a *non-mutating* version of quicksort; it does not mutate the
       input list.
                                               Lists of length < 2 are
       if len(lst) < 2:
                                                   already sorted
           return |st[:]
       else:
                                                                                Simple partition step
           # smaller, bigger = partition(lst[1:], lst[0])
           smaller = [i \text{ for } i \text{ in } lst[1:] \text{ if } i < lst[0]]
           bigger = [i for i in lst[1:] if i >= lst[0]]
                                                                      Sort smaller elements
           return (quicksort(smaller) +
                    [lst[o]] + \_
                    quicksort(bigger) )
                                                                       in its correct position
Sort larger elements
```



Counting Quicksort: n = 7





```
def quicksort(lst):
   if len(lst) < 2:
      return lst[:]
   else:
      pivot = lst[0]
      smaller, bigger = partition(lst[1:], pivot)
      smaller sorted = quicksort(smaller)
      bigger sorted = quicksort(bigger)
      return smaller sorted + [pivot] +
             bigger sorted
```



Do we always have *n log n*?

Mergesort: we know we always split in halves, no matter what

Quicksort: no guarantees, depends on how we pick the pivot

What's the average case? What's the worst case?



Quicksort: Good on Average

 If we always choose a pivot that's an approximate median, then the two partitions are roughly equal, and the running time is O(n log(n))

 If we always choose a pivot that's an approximate min/max, then they two partitions are very unequal, and the running time is O(n²)

 In practice, with good pivot selection, it's rare to come across worst case behaviour.



The limitations of Big-Oh

- Big-Oh notation is a simplification of running time analysis and allows us to ignore constants when analysing efficiency.
- But constants can make a difference, too!
- O(n log n) Merge Sort vs. O(n log n) Quicksort vs. O(n²) Bubble