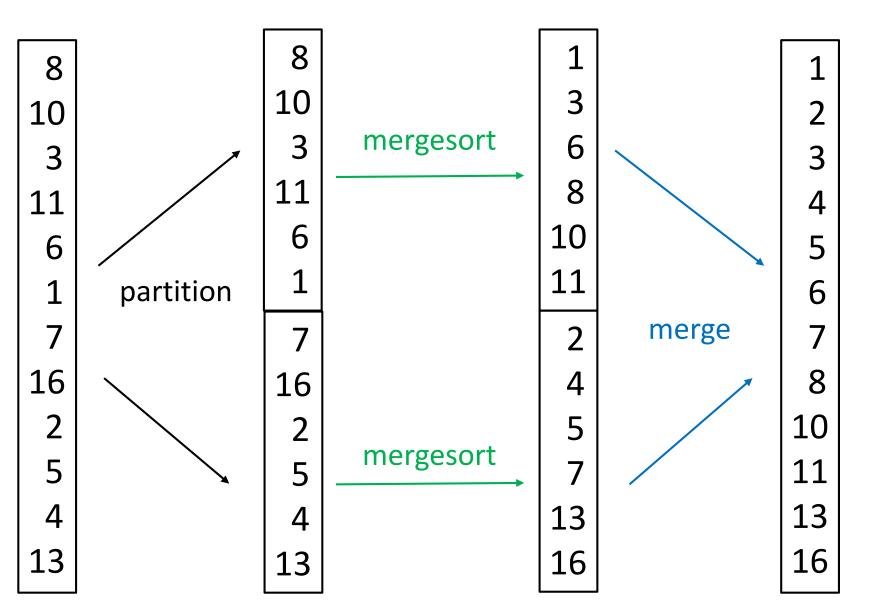
COMP 250

Lecture 34

recurrences 2: mergesort & quicksort

Nov. 28, 2018

Example 5: Mergesort



```
mergesort(list){
  if list.length == 1
     return list
  else{
     mid = (list.size - 1) / 2
     list1 = list.getElements(0,mid)
     list2 = list.getElements(mid+1, list.size-1)
     list1 = mergesort(list1)
     list2 = mergesort (list2)
     return merge(list1, list2)
          t(n) = c n + 2 t \left(\frac{n}{2}\right)
```

We are ignoring a constant term for simplicity.

```
mergesort(list){
   if list.length == 1
                                     Base case n = 1
     return list
  else{
                                                        getElements could take
     mid = (list.size - 1) / 2
                                                        time proportional to n.
     list1 = list.getElements(0,mid)
     list2 = list.getElements(mid+1, list.size-1)
     list1 = mergesort(list1)
     list2 = mergesort (list2)
     return merge(list1, list2)
          t(n) = c n + 2 t \left(\frac{n}{2}\right)
                                                               We are ignoring a
                                                               constant term for
```

simplicity.

Recall same issue with binary search last lecture:

What if *n* is not even ?

e.g.
$$t(13) = c * 13 + t(6) + t(7)$$

In general, one should write the recurrence as:

$$t(n) = c n + t \left(floor\left(\frac{n}{2}\right)\right) + t \left(ceiling\left(\frac{n}{2}\right)\right)$$
round down round up

In COMP 250, one typically assumes $n=2^k$ for recurrences that involve $t(\frac{n}{2})$. The more general recurrence has roughly the same solution.

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$= c n + 2 (c\frac{n}{2} + 2t(\frac{n}{4}))$$

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$= c n + 2 (c\frac{n}{2} + 2t(\frac{n}{4}))$$

$$= c n + c n + 4 t(\frac{n}{4})$$

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$= c n + 2 (c\frac{n}{2} + 2t(\frac{n}{4}))$$

$$= c n + c n + 4 t(\frac{n}{4})$$

$$= c n + c n + 4 (c\frac{n}{4} + 2 t(\frac{n}{8}))$$

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$= c n + 2 (c\frac{n}{2} + 2t(\frac{n}{4}))$$

$$= c n + c n + 4 t(\frac{n}{4})$$

$$= c n + c n + 4 (c\frac{n}{4} + 2 t(\frac{n}{8}))$$

$$= c n + c n + c n + 8 t(\frac{n}{8})$$

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$= c n + 2 (c\frac{n}{2} + 2t(\frac{n}{4}))$$

$$= c n + c n + 4 t(\frac{n}{4})$$

$$= c n + c n + 4 (c\frac{n}{4} + 2 t(\frac{n}{8}))$$

$$= c n + c n + c n + 8 t(\frac{n}{8})$$

$$= c n k + 2^k t(\frac{n}{2^k})$$

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$= c n + 2 (c\frac{n}{2} + 2t(\frac{n}{4}))$$

$$= c n + c n + 4 t(\frac{n}{4})$$

$$= c n + c n + 4 (c\frac{n}{4} + 2 t(\frac{n}{8}))$$

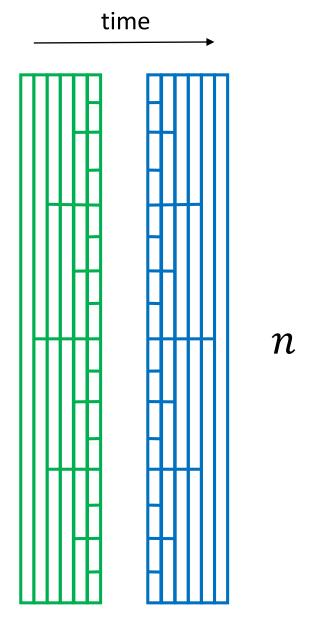
$$= c n + c n + c n + 8 t(\frac{n}{8})$$

$$= c n k + 2^k t(\frac{n}{2^k})$$

$$= c n \log_2 n + n t(1), \text{ when } n = 2^k$$

```
mergesort(list){
  if list.length == 1
    return list
  else{
    mid = (list.size - 1) / 2
    list1 = list.getElements(0,mid)
    list2 = list.getElements(mid+1, list.size-1)
    list1 = mergesort(list1)
    list2 = mergesort (list2)
    return merge( list1, list2 )
  }
}
```

Q: How many recursive calls are made to mergesort?



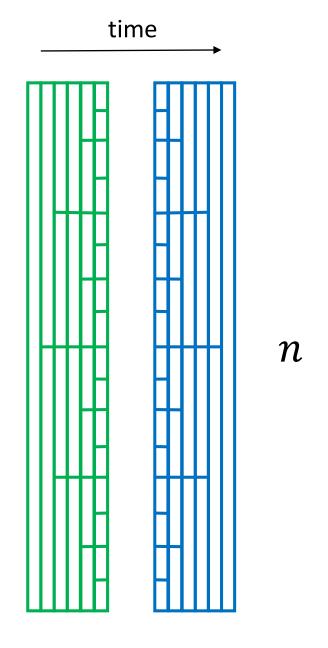
n

 log_2n log_2n

```
mergesort(list){
  if list.length == 1
    return list
  else{
    mid = (list.size - 1) / 2
    list1 = list.getElements(0,mid)
    list2 = list.getElements(mid+1, list.size-1)
    list1 = mergesort(list1)
    list2 = mergesort (list2)
    return merge( list1, list2 )
  }
}
```

Q: How many recursive calls are made to mergesort?

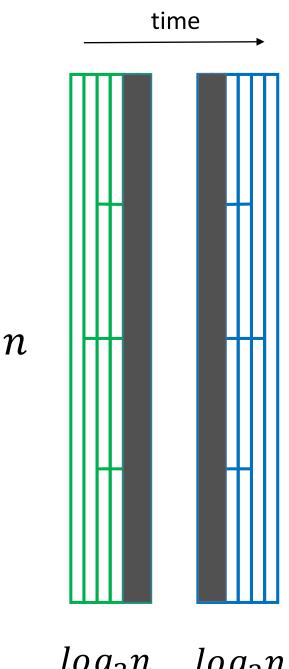
A: n - 1.



n

 log_2n log_2n

What if we change the base case and stop the recursion at a larger list size $n_0 > 1$?



For example, if we stop at $n_0 > 4$, then we don't have to do the recursions in the grayed out part.

But we still need to sort and merge.

How does that change the total time?

 log_2n log_2n

```
mergesort(list){
  if list.length < 5
                                 // or some other base case
     return bubblesort(list)
  else{
     mid = (list.size - 1) / 2
     list1 = list.getElements(0,mid)
     list2 = list.getElements(mid+1, list.size-1)
     list1 = mergesort(list1)
     list2 = mergesort (list2)
     return merge(list1, list2)
```

$$t(n) = c n + 2 t \left(\frac{n}{2}\right), \qquad n \ge 5$$

How does this affect the back substitution and solution?

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$t(n) = c n + 2 t(\frac{n}{2})$$

$$= \dots$$

$$= c n k + 2^k t(\frac{n}{2^k})$$

Stop back substitution when $\frac{n}{2^k} = 4$

$$t(n) = c n + 2 t(\frac{n}{2})$$

= ...
= $c n k + 2^k t(\frac{n}{2^k})$, and letting $2^k = \frac{n}{4}$ gives...
= $c n(\log_2 n - 2) + \frac{n}{4} t(4)$

$$t(n) = c n + 2 t(\frac{n}{2})$$

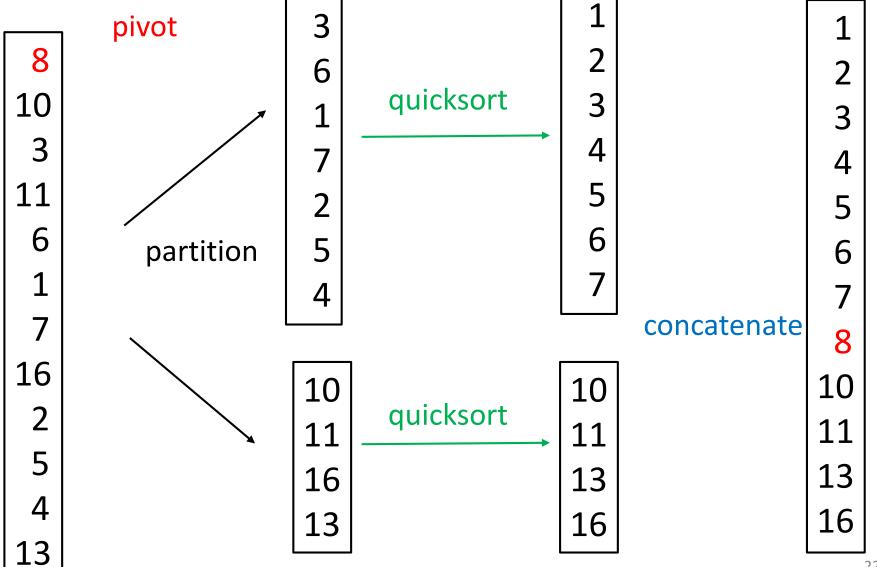
$$= \dots$$

$$= c n k + 2^k t(\frac{n}{2^k}), \text{ and letting } 2^k = \frac{n}{4} \text{ gives...}$$

$$= c n(\log_2 n - 2) + \frac{n}{4} t(4)$$

For large n, the dominant term is still $n \log_2 n$.

Quicksort



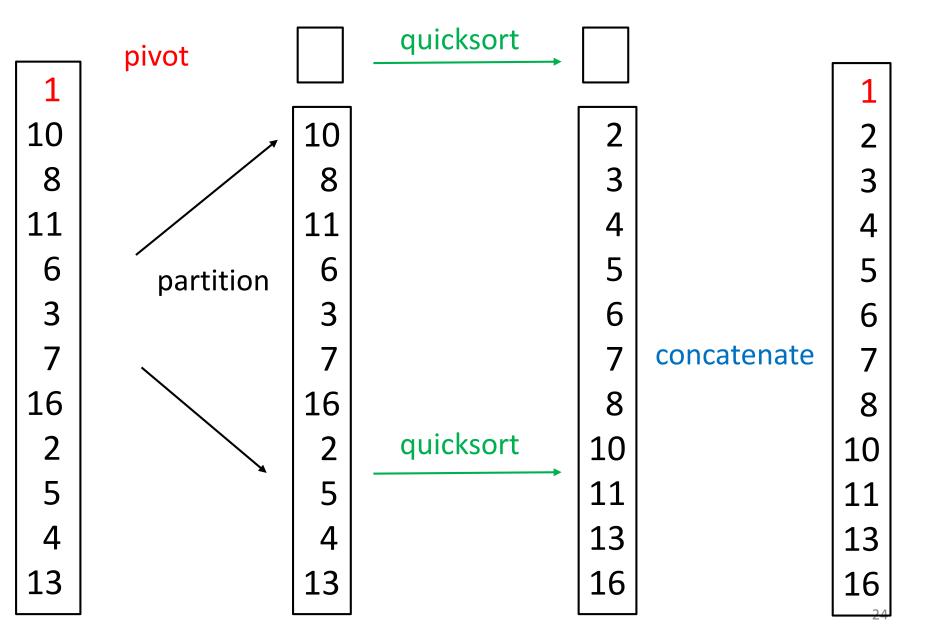
Quicksort Best Case

$$t(n) = c n + 2 t \left(\frac{n}{2}\right)$$
Partition based on the pivot

Two sublists have the same size

This is the same recurrence as mergesort.

Quicksort worst case?



Quicksort worst case

$$t(n) = c n + t(n-1)$$

From last lecture....

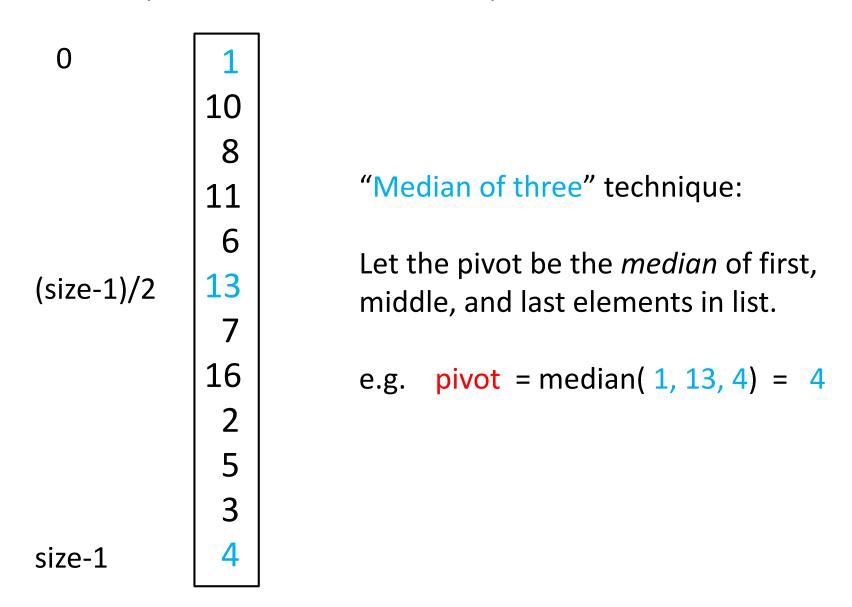
Quicksort worst case

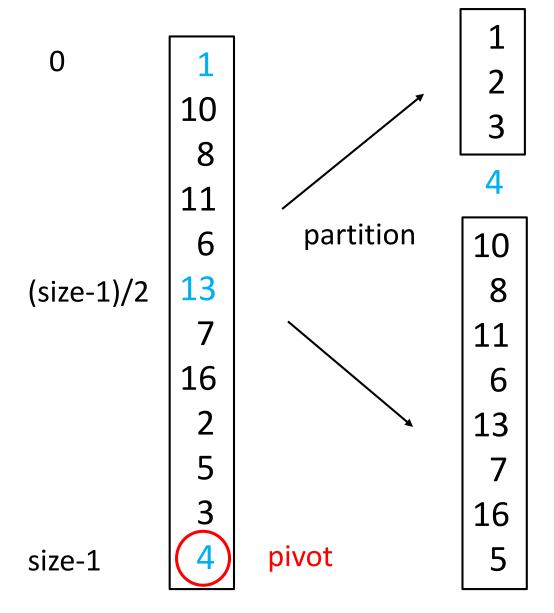
$$t(n) = c n + t(n-1)$$

From last lecture.... (algorithm was similar to selection sort)

$$t(n) = c \frac{n(n+1)}{2}$$

How to reduce the chance of an unbalanced partition at each step of Quicksort?





Works well in practice, especially if the list is already nearly sorted.

Example 6

$$t(n) = n + t(\frac{n}{2})$$

What do you expect?

(BTW, using cn for the first term instead of n won't change the following derivation.)

$$t(n) = n + t(\frac{n}{2})$$
$$= n + \frac{n}{2} + t(\frac{n}{4})$$

$$t(n) = n + t(\frac{n}{2})$$

$$= n + \frac{n}{2} + t(\frac{n}{4})$$

$$= n + \frac{n}{2} + \frac{n}{4} + t(\frac{n}{8})$$

$$t(n) = n + t(\frac{n}{2})$$

$$= n + \frac{n}{2} + t(\frac{n}{4})$$

$$= n + \frac{n}{2} + \frac{n}{4} + t(\frac{n}{8})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + \frac{n}{2^{k-1}} + t(\frac{n}{2^k})$$

$$t(n) = n + t(\frac{n}{2})$$

$$= n + \frac{n}{2} + t(\frac{n}{4})$$

$$= n + \frac{n}{2} + \frac{n}{4} + t(\frac{n}{8})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + \frac{n}{2^{k-1}} + t(\frac{n}{2^k})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + t(1), \text{ when } 2^k = n$$

$$t(n) = n + t(\frac{n}{2})$$

$$= n + \frac{n}{2} + t(\frac{n}{4})$$

$$= n + \frac{n}{2} + \frac{n}{4} + t(\frac{n}{8})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + \frac{n}{2^{k-1}} + t(\frac{n}{2^k})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + t(1), \text{ when } 2^k = n$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + 1 - 1 + t(1)$$

$$t(n) = n + t(\frac{n}{2})$$

$$= n + \frac{n}{2} + t(\frac{n}{4})$$

$$= n + \frac{n}{2} + \frac{n}{4} + t(\frac{n}{8})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + \frac{n}{2^{k-1}} + t(\frac{n}{2^k})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + t(1), \quad \text{when } 2^k = n$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + 1 - 1 + t(1)$$

$$\log_2 n$$

$$\sum_{i = 1}^{\log_2 n} 2^i = ?$$

$$t(n) = n + t(\frac{n}{2})$$

$$= n + \frac{n}{2} + t(\frac{n}{4})$$

$$= n + \frac{n}{2} + \frac{n}{4} + t(\frac{n}{8})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + \frac{n}{2^{k-1}} + t(\frac{n}{2^k})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + t(1), \text{ when } 2^k = n$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + 1 - 1 + t(1)$$

$$\log_2 n$$

$$\sum_{i = 1}^{\log_2 n} 2^i = 2^{\log_2 n + 1} - 1 = ?$$

$$t(n) = n + t(\frac{n}{2})$$

$$= n + \frac{n}{2} + t(\frac{n}{4})$$

$$= n + \frac{n}{2} + \frac{n}{4} + t(\frac{n}{8})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + \frac{n}{2^{k-1}} + t(\frac{n}{2^k})$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + t(1), \text{ when } 2^k = n$$

$$= n + \frac{n}{2} + \frac{n}{4} + \dots + 4 + 2 + 1 - 1 + t(1)$$

$$\log_2 n$$

$$\sum_{i = 2^{log_2 n + 1}} 2^i = 2^{log_2 n + 1} - 1 = 2n - 1$$

We have solved the following recurrences:

$$t(n) = c + t(n-1)$$

$$t(n) = c n + t(n-1)$$

$$t(n) = c + 2t(n-1)$$

$$t(n) = c + t(\frac{n}{2})$$

$$t(n) = c n + 2 t(\frac{n}{2})$$

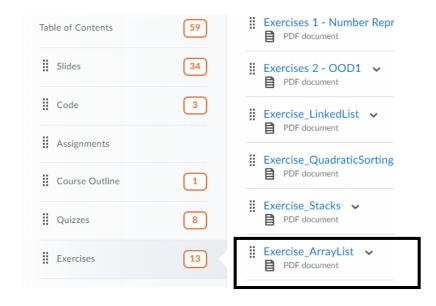
$$t(n) = c n + t(\frac{n}{2})$$

In COMP 251, you will prove a general result called the Master Theorem which covers all these cases and more!

$$1 + 2 + 4 + \dots + 2^{\log_2 n} = 2n - 1$$

$$1 + 2 + 4 + \dots + 2^{\log_2 n} = 2n - 1$$

• Adding n elements to an empty arraylist and resizing log_2n times.



See Q6 for details.

$$1 + 2 + 4 + \dots + 2^{\log_2 n} = 2n - 1$$

- Adding n elements to an empty arraylist.
- Putting n elements into an empty hash table.
 (The rehashing takes the same amount of time as in the array list example. Why?)

$$1 + 2 + 4 + \dots + 2^{\log_2 n} = 2n - 1$$

- Adding n elements to an empty arraylist.
- Putting n elements into an empty hash table.

So, on average, each of these operations is O(1).

In COMP 251, this is called "amortized" analysis.

TODO

• lecture 35, 36, 37: Asymptotic Complexity

Quiz 5 on Friday Nov 30

Room change on Friday Nov 30 for Sec. 001.
 (10:35-11:25) ENGTR 0100