

Lecture 06: Sorting by Divide and Conquer II

COSC 311 *Algorithms*, Fall 2022

Announcement

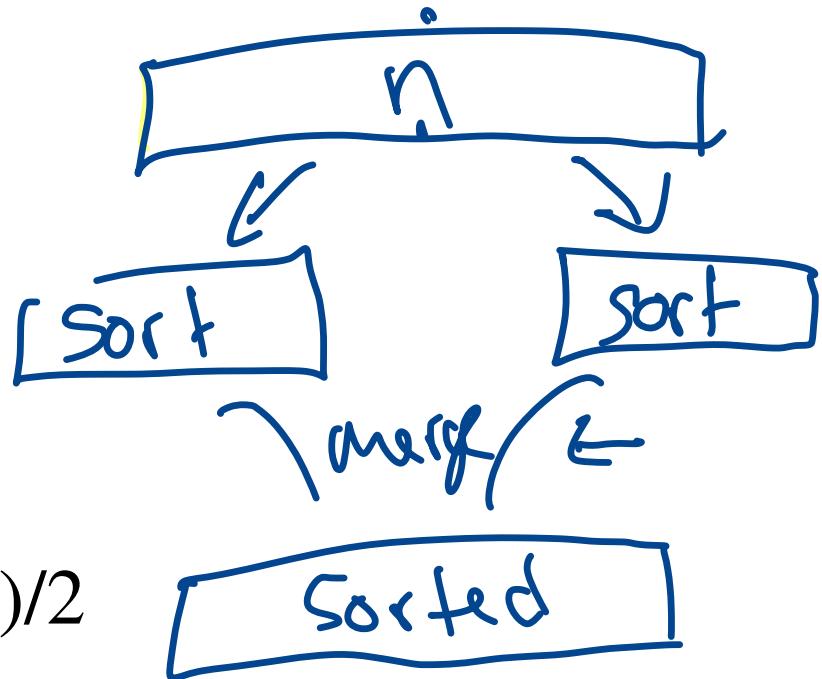
Accountability Groups

Overview

1. MergeSort
2. Running Time of Merge Sort
3. QuickSort

Previously

- Sorting meets Divide & Conquer
- MergeSort: Divide by Index
 1. divide a into halves $m = (n + 1)/2$
 2. sort $a[1..m - 1]$ recursively
 3. sort $a[m..n]$ recursively
 4. merge $a[1..m - 1]$ and $a[m..n]$ to form sorted array

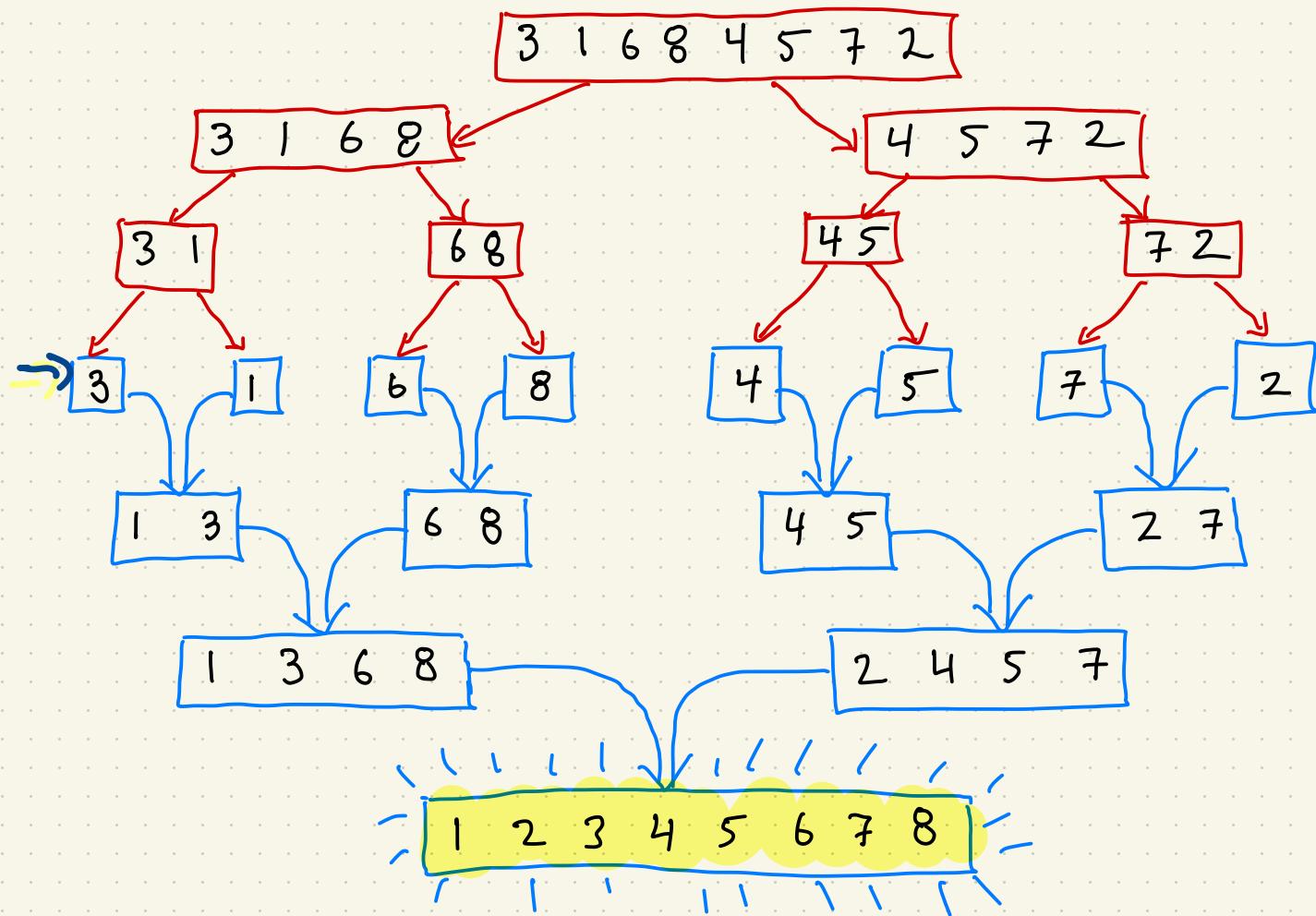


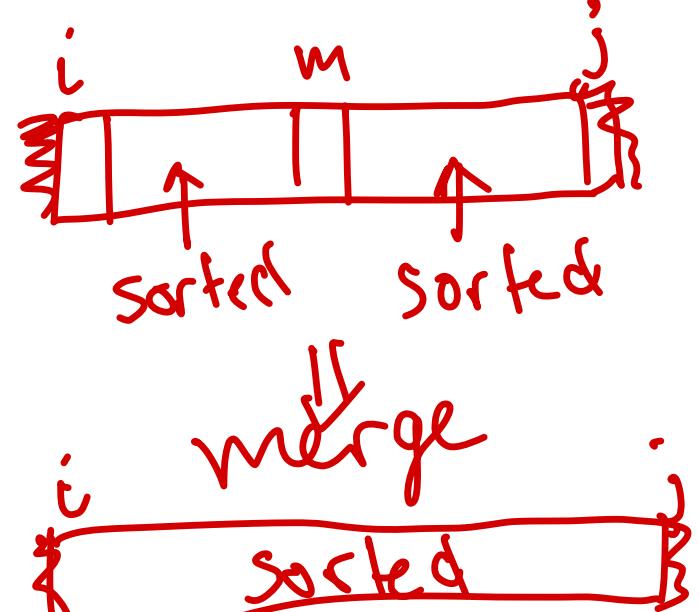
Pseudocode

```
# sort values of a between indices i and j-1
MergeSort(a, i, j):
    if j - i = 1 then
        return
    endif
    m <- (i + j) / 2
    MergeSort(a, i, m)
    MergeSort(a, m, j)
    Merge(a, i, m, j)
```

base case
sort left half
sort right half
merge halves

Illustration of MergeSort





Correctness of MergeSort

Establish two claims:

Claim 1 (merge). If $a[i..m - 1]$ and $a[m..j]$ are sorted, then after $\text{Merge}(a, i, m, j)$, $a[i..j]$ is sorted.

- Argued on lecture ticket!

Claim 2. For any indices $i < j$, after calling $\text{MergeSort}(a, i, j)$, $a[i..j]$ is sorted.

- Argue by Induction!

Pseudocode Again

```
00 # sort values of a between indices i and j-1
01 MergeSort(a, i, j).
02 if j - i = 1 then
03     return
04 endif
05 m <- (i + j) / 2
06 MergeSort(a, i, m)
07 MergeSort(a, m, j)
08 Merge(a, i, m, j)
```

$k=1$ just return all arrays of size 1
and sorted!

} ← have size $\frac{k+1}{2} < k$ ←

Output is sorted by claim 1

succeed by ind. hyp.

Inductive Claim

of elements
↓ being sorted

Consider $\text{MergeSort}(a, i, j)$, define $k = j - i$ to be size

$P(k)$: for every $k' \leq k$, $\text{MergeSort}(a, i, j)$ with size k' succeeds

Base case $k = 1$:

All arrays of size 1 are sorted!

Inductive step $P(k) \Rightarrow P(k + 1)$:

Assume MergeSort sorts all arrays of size up to k .

Show Sorts arrays of size $k+1$:

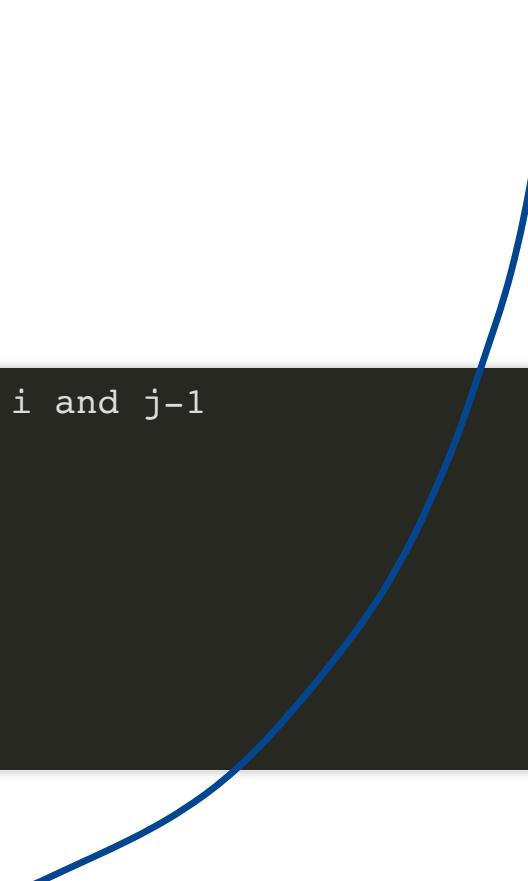
Selection Sort : $\Theta(n^2)$

Question

How efficient is MergeSort?

```
00 # sort values of a between indices i and j-1
01 MergeSort(a, i, j):
02     if j - i = 1 then
03         return
04     endif
05     m <- (i + j) / 2
06     MergeSort(a,i,m)
07     MergeSort(a,m,j)
08     Merge(a,i,m,j)
```

$\Theta(n \log n)$



Analyzing Running Time

```
00 # sort values of a between indices i and j-1
01 MergeSort(a, i, j):
02     if j - i = 1 then
03         return
04     endif
05     m <- (i + j) / 2
06     MergeSort(a, i, m)
07     MergeSort(a, m, j)
08     Merge(a, i, m, j)
```

$O(1)$

??
..

$O(k)$

Observation 1. Let $k = j - i$ be the size of the method call $\text{MergeSort}(a, i, j)$. Then running time is $O(k) +$ running time of recursive calls on lines 6-7.

Observation 2. Recursive calls have size $k/2$.

- Assume size is power of 2

Combining Observations

Looking @ recursive calls by depth:

Calls (size)

(n)

no. calls

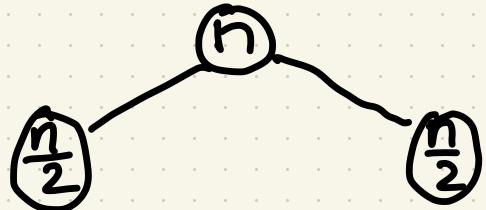
1
~

size of calls

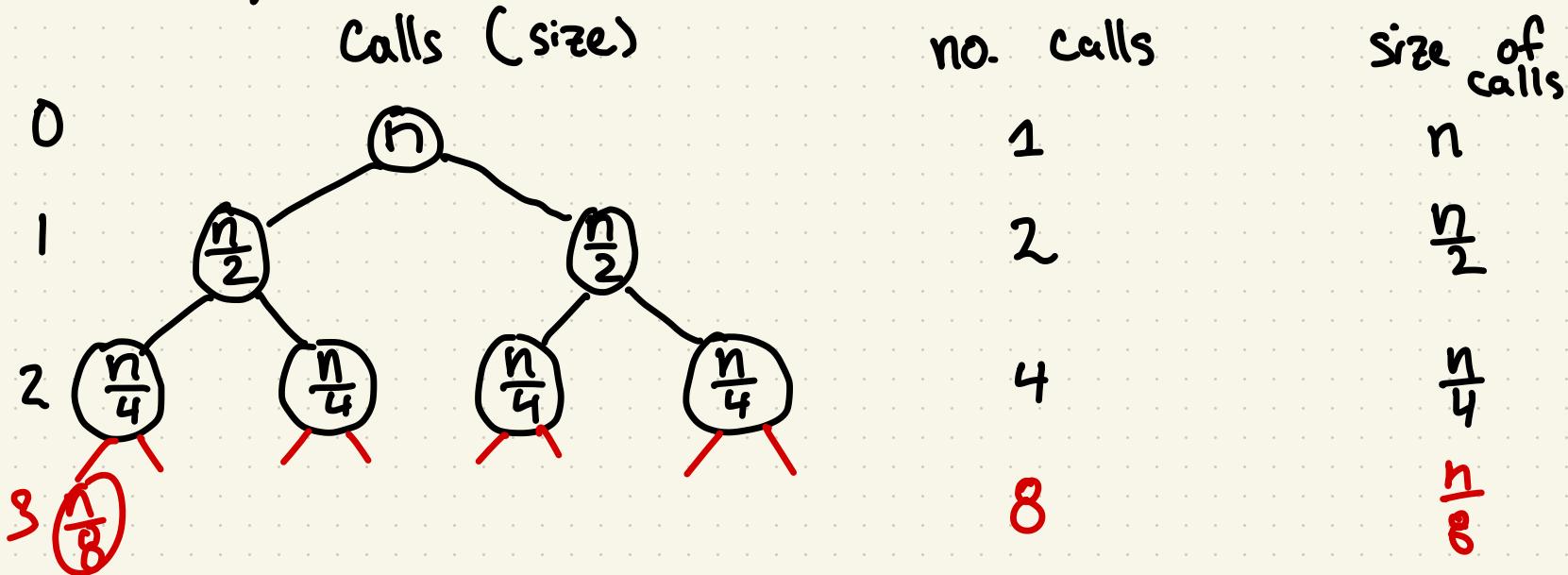
n

Looking @ recursive calls by depth:

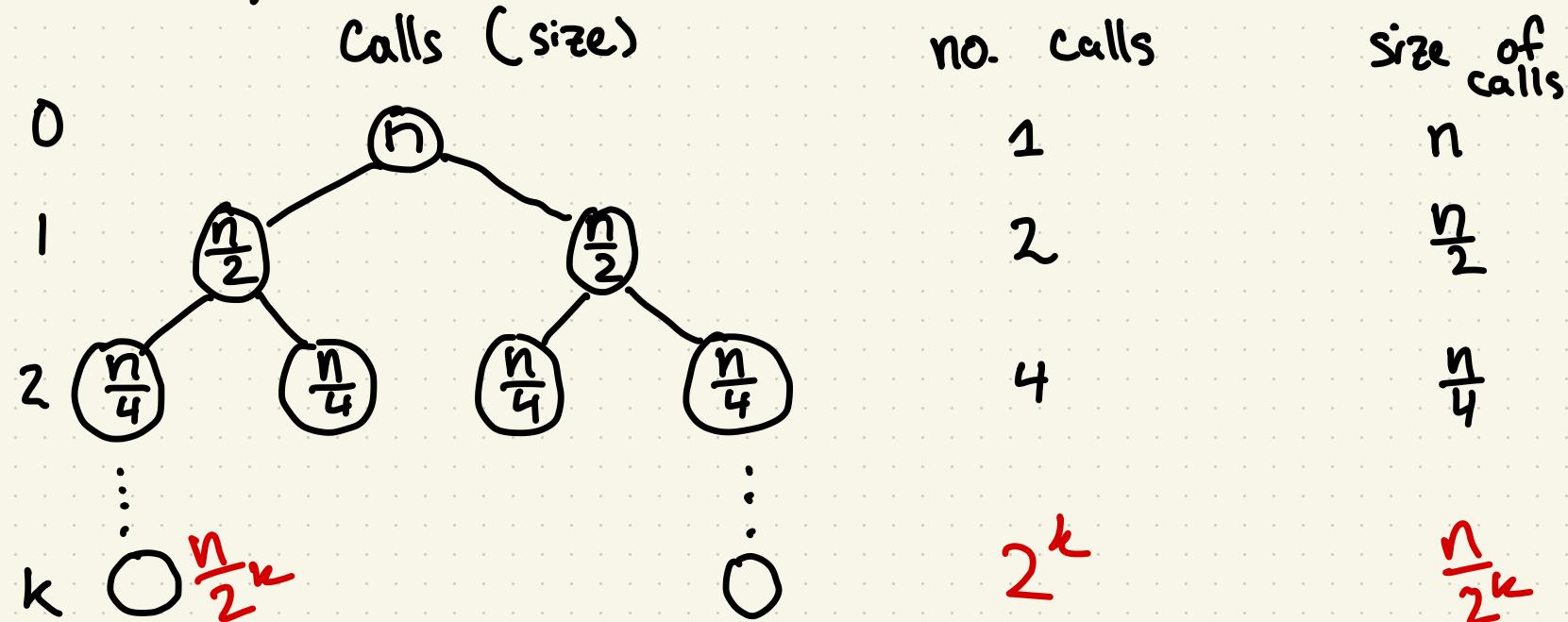
Calls (size)	no. calls	size of calls
n	1	n
$\frac{n}{2}$	2	$\frac{n}{2}$



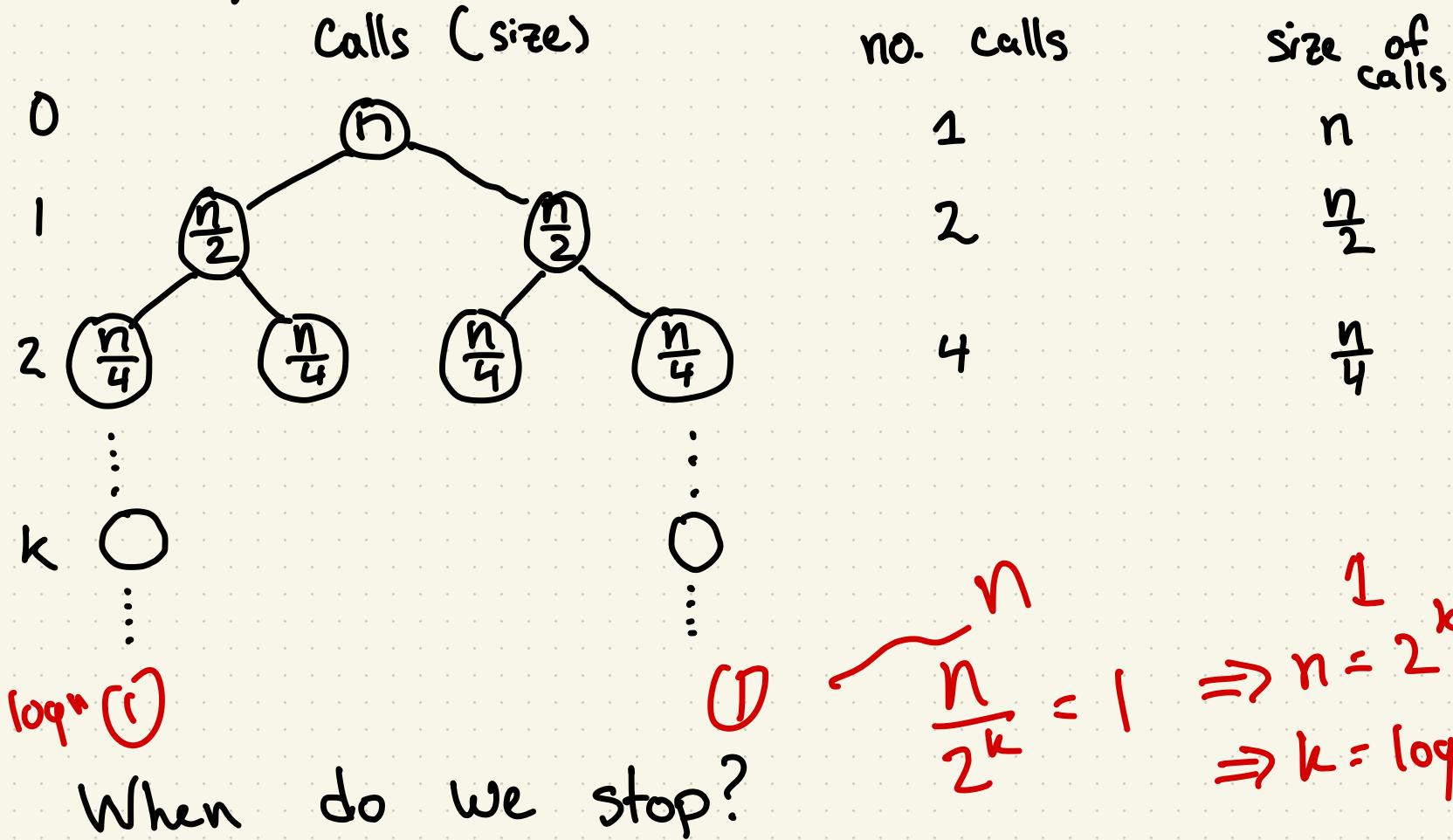
Looking @ recursive calls by depth:



Looking @ recursive calls by depth:



Looking @ recursive calls by depth:



$$\frac{n}{2^k} = 1 \Rightarrow n = 2^k$$
$$\Rightarrow k = \log n$$

Recall Logarithms (base 2)

Define \log by

- $\log a = b \iff 2^b = a$

Another way

- $\log a$ is # times a can be divided by 2 to get (at most) 1.

Facts.

1. For every constant $c > 0$, $\log n = O(n^c)$.
2. $\log n \neq O(1)$.

Running Times by Depth

merge, etc.

depth	no.	size	time (not counting rec. calls)
0	1	n	$1 \cdot \Theta(n) = \Theta(n)$
1	2	$n/2$	$2 \cdot \Theta(n/2) = \Theta(n)$
2	4	$n/4$	$4 \cdot \Theta(n/4) = \Theta(n)$
:	:		
k	2^k	$n/2^k$	$2^k \cdot \Theta(n/2^k) = \Theta(n)$
:	:		
$\log n$	n	1	$n \cdot \Theta(1) = \Theta(n)$

$\Theta(n \log n)$

$\log n$ depth
 \Rightarrow total r.t. $(\log n) \cdot \Theta(n)$

Foresighting

$T(n)$ = running time of MergeSort on α of size n
Worst-case

$T(n)$ satisfies the recursion relation merge,
etc.

$$T(n) = 2 \cdot T(n/2) + \Theta(n)$$

"Master Method" gives gen. formula for this
type of recursion

recursive calls

$$\Rightarrow T(n) = \Theta(n \log n)$$

Picture so Far:

SelectionSort. $O(n^2)$ operations

- $O(n^2)$ comparisons
- $O(n)$ swaps



BubbleSort and InsertionSort. $O(n^2)$ operations

- $O(n^2)$ comparisons
- $O(n^2)$ swaps

MergeSort. $O(n \log n)$ operations

- $O(n \log n)$ comparisons
- $O(n \log n)$ modifications
- uses $O(n)$ space overhead