

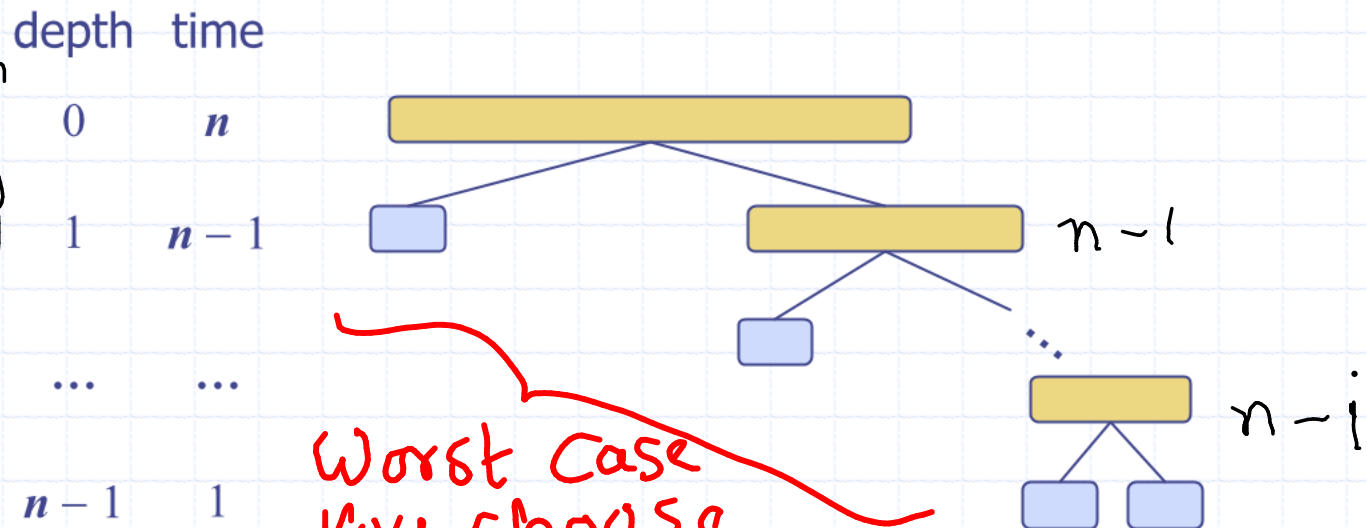
Worst-case Running Time

- ◆ The worst case for quick-sort occurs when the pivot is the unique minimum or maximum element
- ◆ One of L and G has size $n - 1$ and the other has size 0
- ◆ The running time is proportional to the sum

$$n + (n - 1) + \dots + 2 + 1 = O(n^2)$$

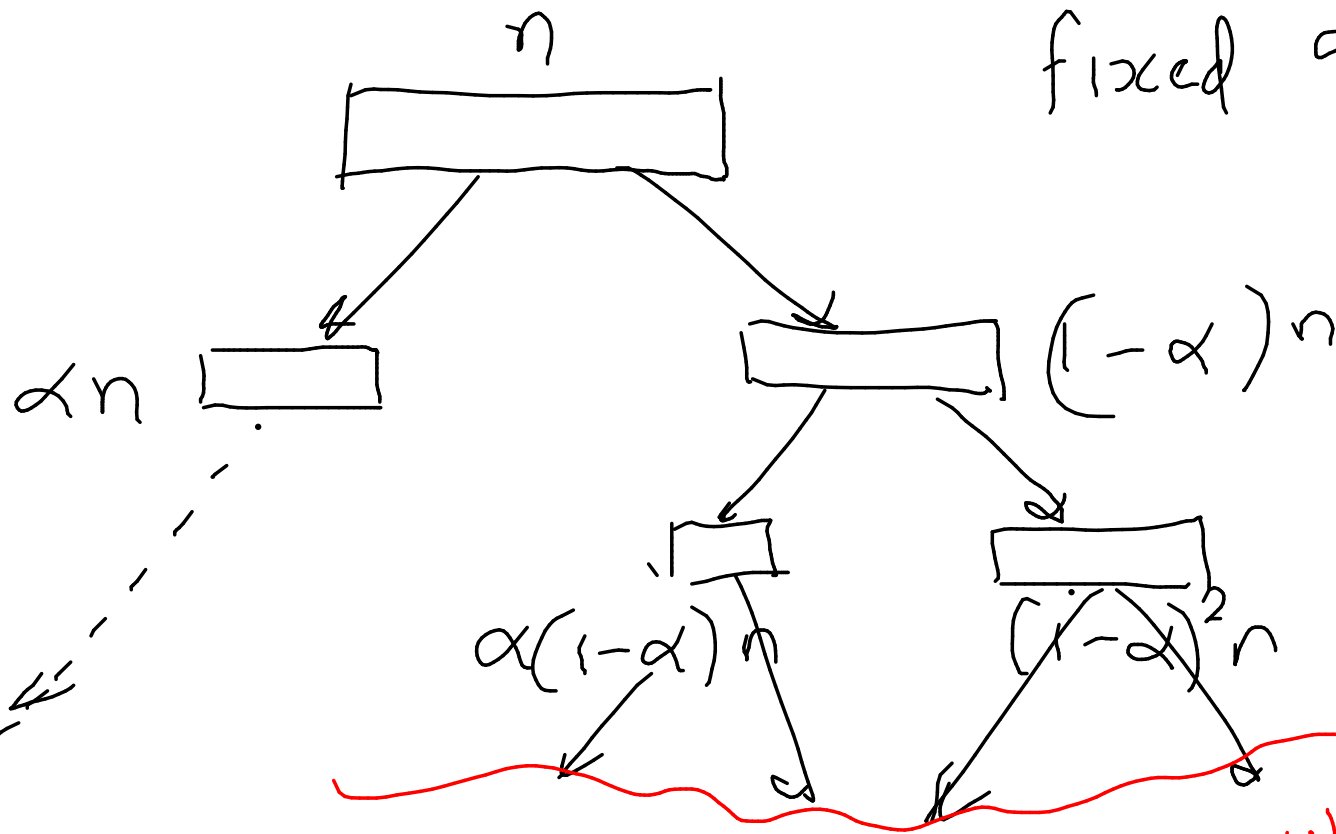
- ◆ Thus, the worst-case running time of quick-sort is $O(n^2)$

For worst case, functions like selection sort



Worst Case
you choose
Smallest element as
pivot at each step

fixed $\alpha \in (0, 1)$



will end early

$\alpha^i n = 1$

If wlog $\alpha > 0.5$

then longest branch will have

$$\alpha^h n = 1 \Rightarrow h = O(\log_{1/\alpha} n) \\ = O(\log_2 n)$$

without loss of generality

You can prove that Quicksort will be $O(n \log n)$ in this case

Thus: Most Qsort partitions
are not bad.

Let us consider amortized
analysis of Qsort

Assume: Fixed array of length n
we will average over all choices
of pivot

Let $T(n)$ = average time for quicksort, average
over all choices (n in number) of
pivot

Recurrence relation:

$$T(n) = \frac{1}{n} \left[\begin{array}{l} \underbrace{T(n-1) + T(0) + n} \\ + T(n-2) + T(1) + n \\ \vdots \\ + \underbrace{T(n-k-1) + T(k) + n} \\ \vdots \\ + T(0) + T(n-1) + n \end{array} \right]$$

Note:

→ ① n is reqd for swaps

② The pivot element need not be further analysed
 $(n-k-1) + k = n-1$

$$T(n) = \left(\frac{2}{n} \sum_{i=0}^{n-1} T(i) \right) + n \rightarrow \textcircled{A}$$

$$T(n-1) = \left(\frac{2}{n-1} \sum_{i=0}^{n-2} T(i) \right) + n-1 \rightarrow \textcircled{B}$$

Rescaling (B) and substituting in (A)

$$T(n) = \frac{2}{n} T(n-1) + \frac{n-1}{n} T(n-1) - \frac{(n-1)^2}{n} + n$$

$$= \frac{n+1}{n} T(n-1) + \left(2 - \frac{1}{n}\right) \rightarrow \textcircled{C}$$

Replacing
n with
n-1

$$T(n-1) = \frac{n}{n-1} T(n-2) + \left(2 - \frac{1}{n-1}\right) \rightarrow \textcircled{D}$$

Substituting for $T(n-1)$ from (D) into (C)

$$T(n) = \frac{n+1}{n-1} T(n-2) + \frac{n+1}{n} \left(2 - \frac{1}{n-1}\right) + \left(2 - \frac{1}{n}\right)$$

$$= \frac{n+1}{n-1} T(n-2) + \frac{2n}{n} + \frac{2}{n} - \frac{n+1}{n(n-1)} + 2 - \frac{1}{n}$$

$$= \frac{n+1}{n-1} T(n-2) + \frac{2n}{n} + \frac{1}{n} - \frac{n+1}{n^2-n} + 2$$

$$\vdots = \frac{n+1}{n-i+1} \cdot T(n-i) + 2(n+1) \left(\frac{1}{n} + \frac{1}{n-1} + \dots + \frac{1}{n-i+1} \right)$$

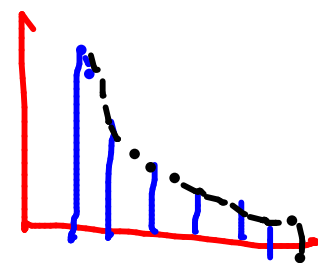
$$= \frac{n+1}{1} T(0) + 2(n+1) \left(\frac{1}{n} + \dots + \frac{1}{2} + 1 \right) + 2$$

\downarrow
 $T(0) = 0$

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

$$\leq \left(2(n+1) \int_1^n \frac{1}{x} dx \right) + 2$$

(Area below curve is \geq Sum of blue bars)



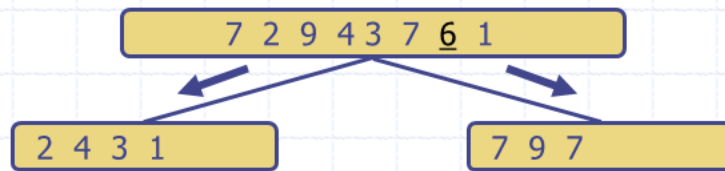
$$= 2(n+1) \log n + 2$$

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

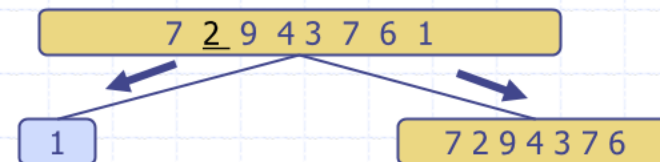
You could also substitute $T(n) = O(n \log n)$
 into (C) & prove that they are
 consistent by using recurrence

(Optional Reading: Another loose way of Expected Running Time amortized analysis)

- ◆ Consider a recursive call of quick-sort on a sequence of size s
 - **Good call:** the sizes of L and G are each less than $3s/4$
 - **Bad call:** one of L and G has size greater than $3s/4$



Good call



Bad call

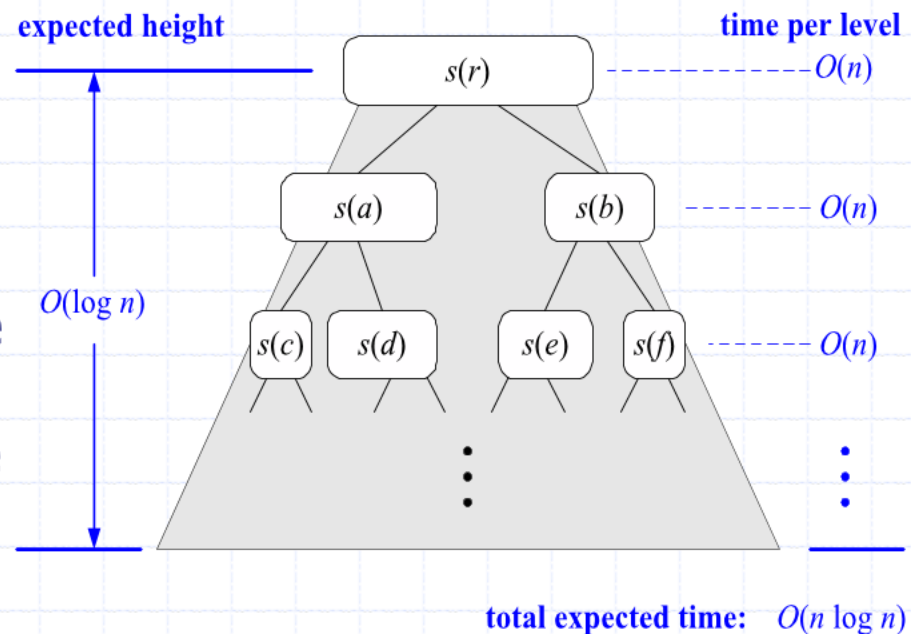
- ◆ A call is **good** with probability $1/2$
 - $1/2$ of the possible pivots cause good calls:



(Optional reading)

Expected Running Time, Part 2

- ◆ **Probabilistic Fact:** The expected number of coin tosses required in order to get k heads is $2k$
- ◆ For a node of depth i , we expect
 - $i/2$ ancestors are good calls
 - The size of the input sequence for the current call is at most $(3/4)^{i/2}n$
- ◆ Therefore, we have
 - For a node of depth $2\log_{4/3}n$, the expected input size is one
 - The expected height of the quick-sort tree is $O(\log n)$
- ◆ The amount of work done at the nodes of the same depth is $O(n)$
- ◆ Thus, the expected running time of quick-sort is $O(n \log n)$



Summary of Sorting Algorithms

Algorithm	Time	Notes
selection-sort	$O(n^2)$	<ul style="list-style-type: none">◆ in-place◆ slow (good for small inputs)
insertion-sort	$O(n^2)$	<ul style="list-style-type: none">◆ in-place◆ slow (good for small inputs)
quick-sort	$O(n \log n)$ expected	<ul style="list-style-type: none">◆ in-place, randomized◆ fastest (good for large inputs)
heap-sort	$O(n \log n)$	<ul style="list-style-type: none">◆ in-place◆ fast (good for large inputs)
merge-sort	$O(n \log n)$	<ul style="list-style-type: none">◆ sequential data access◆ fast (good for huge inputs)

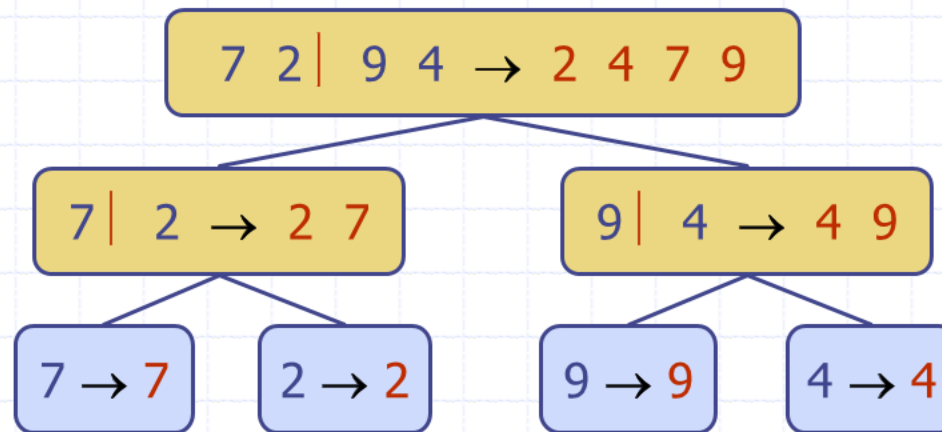
Java Implementation

only works
for distinct
elements

```
public static void quickSort (Object[] S, Comparator c) {
    if (S.length < 2) return; // the array is already sorted in this case
    quickSortStep(S, c, 0, S.length-1); // recursive sort method
}

private static void quickSortStep (Object[] S, Comparator c,
                                   int leftBound, int rightBound) {
    if (leftBound >= rightBound) return; // the indices have crossed
    Object temp; // temp object used for swapping
    Object pivot = S[rightBound];
    int leftIndex = leftBound; // will scan rightward
    int rightIndex = rightBound-1; // will scan leftward
    while (leftIndex <= rightIndex) { // scan right until larger than the pivot
        while ( (leftIndex <= rightIndex) && (c.compare(S[leftIndex], pivot)<=0) )
            leftIndex++;
        // scan leftward to find an element smaller than the pivot
        while ( (rightIndex >= leftIndex) && (c.compare(S[rightIndex], pivot)>=0) )
            rightIndex--;
        if (leftIndex < rightIndex) { // both elements were found
            temp = S[rightIndex];
            S[rightIndex] = S[leftIndex]; // swap these elements
            S[leftIndex] = temp;
        }
    } // the loop continues until the indices cross
    temp = S[rightBound]; // swap pivot with the element at leftIndex
    S[rightBound] = S[leftIndex];
    S[leftIndex] = temp; // the pivot is now at leftIndex, so recurse
    quickSortStep(S, c, leftBound, leftIndex-1);
    quickSortStep(S, c, leftIndex+1, rightBound);
}
```

Divide-and-Conquer



Divide-and-Conquer

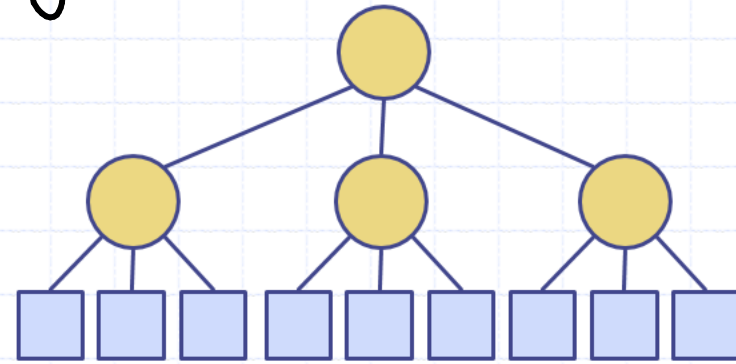
◆ **Divide-and conquer** is a general algorithm design paradigm:

- **Divide**: divide the input data S in two or more disjoint subsets S_1, S_2, \dots
- **Recur**: solve the subproblems recursively
- **Conquer**: combine the solutions for S_1, S_2, \dots , into a solution for S

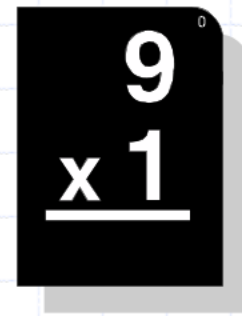
◆ The base case for the recursion are subproblems of constant size

◆ Analysis can be done using **recurrence equations**

Quicksort & mergesort
used only 2 disjoint
subsets



Multiway trees
eg: Red black
& splay



Integer Multiplication

◆ Algorithm: Multiply two n -bit integers I and J .

- Divide step: Split I and J into high-order and low-order bits

$$I = I_h 2^{n/2} + I_l$$

$$J = J_h 2^{n/2} + J_l$$

$$\begin{cases} I = I_h I_l \\ J = J_h J_l \end{cases}$$

- We can then define $I*J$ by multiplying the parts and adding:

$$I * J = (I_h 2^{n/2} + I_l) * (J_h 2^{n/2} + J_l)$$

$$= I_h J_h 2^n + I_h J_l 2^{n/2} + I_l J_h 2^{n/2} + I_l J_l$$

- So, $T(n) = 4T(n/2) + n$.

4 subproblems of size $n/2$ each

Each has length $n/2$

shift by $n/2$ bits

$$5000 = 5 \times 10^3$$

$$I_h \approx 2^{n/2} : I_h \ll n$$

in C/C++/Java

An Improved Integer Multiplication Algorithm



◆ Algorithm: Multiply two n -bit integers I and J .

- Divide step: Split I and J into high-order and low-order bits

$$I = I_h 2^{n/2} + I_l$$

$$J = J_h 2^{n/2} + J_l$$

- Observe that there is a different way to multiply parts:

$$\begin{aligned} I * J &= I_h J_h 2^n + [(I_h - I_l)(J_l - J_h) + I_h J_h + I_l J_l] 2^{n/2} + I_l J_l \\ &= I_h J_h 2^n + [(I_h J_l - I_l J_l - I_h J_h + I_l J_h) + I_h J_h + I_l J_l] 2^{n/2} + I_l J_l \\ &= I_h J_h 2^n + (I_h J_l + I_l J_h) 2^{n/2} + I_l J_l \end{aligned}$$

- So, $T(n) = 3T(n/2) + n$

Recurrence Equation Analysis FOR MERGE-SORT

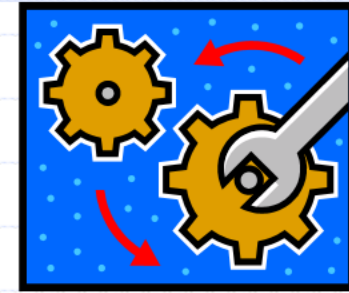


- ◆ The conquer step of merge-sort consists of merging two sorted sequences, each with $n/2$ elements and implemented by means of a doubly linked list, takes at most bn steps, for some constant b .
- ◆ Likewise, the basis case ($n < 2$) will take at b most steps.
- ◆ Therefore, if we let $T(n)$ denote the running time of merge-sort:

$$T(n) = \begin{cases} b & \text{if } n < 2 \\ 2T(n/2) + bn & \text{if } n \geq 2 \end{cases}$$

- ◆ We can therefore analyze the running time of merge-sort by finding a **closed form solution** to the above equation.
 - That is, a solution that has $T(n)$ only on the left-hand side.

Iterative Substitution



- ◆ In the iterative substitution, or “plug-and-chug,” technique, we iteratively apply the recurrence equation to itself and see if we can find a pattern:

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

$$= 2(2T(n/2^2)) + b(n/2) + bn$$

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

$$= 2^3 T(n/2^3) + 3bn$$

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

$$= \dots$$

$$= 2^i T(n/2^i) + ibn$$

- ◆ Note that base, $T(n)=b$, case occurs when $2^i=n$. That is, $i = \log n$.

- ◆ So,

$$T(n) = bn + bn \log n$$

→ dominates

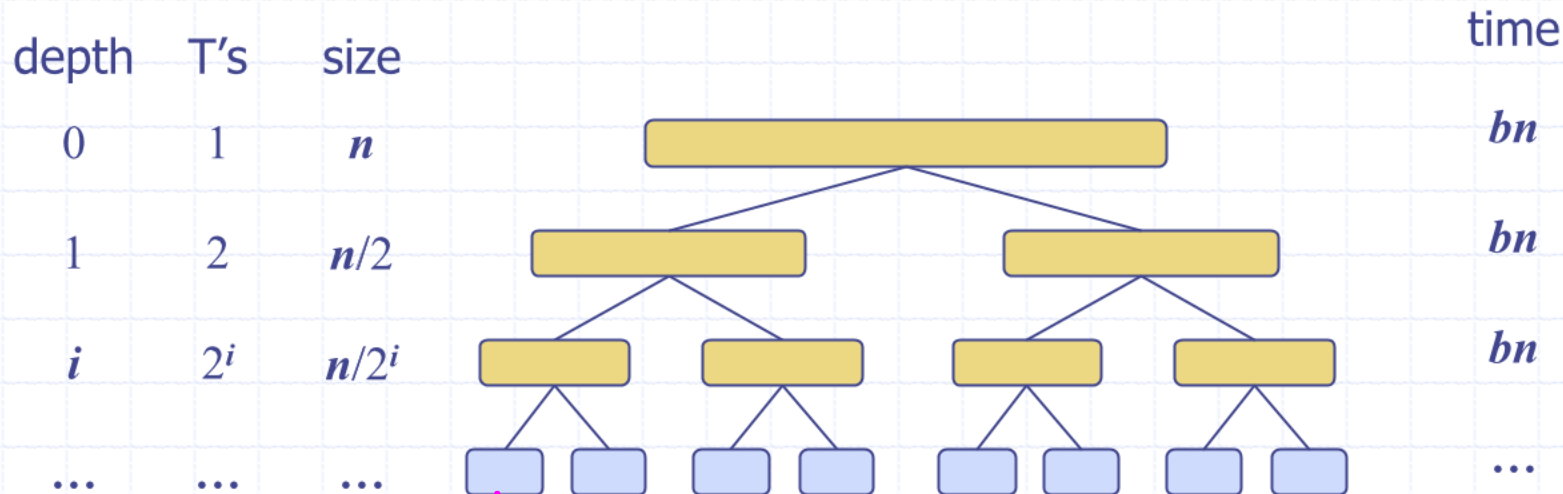
- ◆ Thus, $T(n)$ is $O(n \log n)$.

The Recursion Tree



- ◆ Draw the recursion tree for the recurrence relation and look for a pattern:

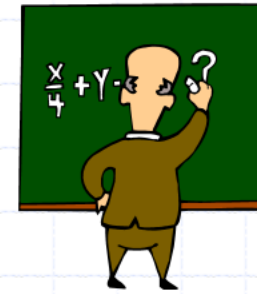
$$T(n) = \begin{cases} b & \text{if } n < 2 \\ 2T(n/2) + bn & \text{if } n \geq 2 \end{cases}$$



b for each leaf node

Total time = $bn + bn \log n$
(last level plus all previous levels)

Guess-and-Test Method



- ◆ In the guess-and-test method, we guess a closed form solution and then try to prove it is true by induction:

$$T(n) = \begin{cases} b & \text{if } n < 2 \\ 2T(n/2) + bn \log n & \text{if } n \geq 2 \end{cases}$$

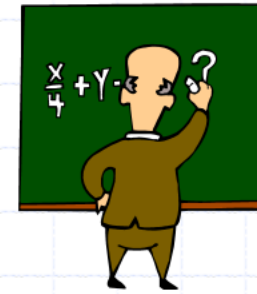
- ◆ Guess: $T(n) < cn \log n$.

$$\begin{aligned} T(n) &= 2T(n/2) + bn \log n \\ &< 2(c(n/2) \log(n/2)) + bn \log n \\ &= cn(\log n - \log 2) + bn \log n \\ &= cn \log n - cn + bn \log n \end{aligned}$$

Note that this recurrence is different from that of mergesort

- ◆ Wrong: we cannot make this last line be less than $cn \log n$

Guess-and-Test Method, Part 2



- ◆ Recall the recurrence equation:

$$T(n) = \begin{cases} b & \text{if } n < 2 \\ 2T(n/2) + bn \log n & \text{if } n \geq 2 \end{cases}$$

- ◆ Guess #2: $T(n) < cn \log^2 n$.

$$T(n) = 2T(n/2) + bn \log n$$

$$= 2(c(n/2) \log^2(n/2)) + bn \log n$$

$$= cn(\log n - \log 2)^2 + bn \log n$$

$$= cn \log^2 n - \underbrace{2cn \log n + cn + bn \log n}_{< 0 \text{ for any } c > b}$$

$$\leq cn \log^2 n$$

- if $c > b$.

- ◆ So, $T(n)$ is $O(n \log^2 n)$.

- ◆ In general, to use this method, you need to have a good guess and you need to be good at induction proofs.

Even $T(n) < n^4$ would have worked!