

EC504 ALGORITHMS AND DATA STRUCTURES
SPRING 2021 MONDAY & WEDNESDAY
2:30 PM - 4:15 PM

Prof: David Castañón, dac@bu.edu

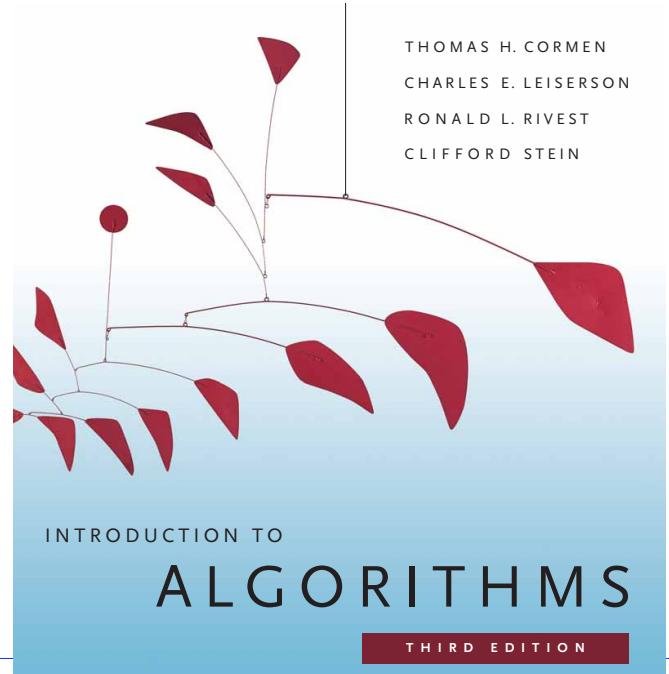
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Course Organization

- **Text:**
 - Cormen, Leiserson, Rivest & Stein (CLRS), Fundamental text
“Introduction to Algorithms” 3rd Edition MIT Pres
- **Reference:**
 - Wikipedia
 - Mark Allen Weiss “Data Structure and Algorithms in C++”.
 - Handouts on specific topics as needed
- **Grading:**
 - HW (Written) 20% ↗
 - HW: Software 20% ↗
 - Project 20% ↗
 - Exams (2) 40% ↗



Course Organization - 2

- Web Site in learn.bu.edu
- On-line classes
- Regular HW (not weekly; 8 or so during term)
- Office hours: TBD, or by appointment. →
- Software: C++, using C++ 11 compatible
- Development: Accounts set up
- HW submission: On Blackboard, with code submitted separately
- Important: This is **not a course to learn software**. We use software to learn about data structures and algorithms, not the other way around... 

Course Organization - 3

- Lectures are recorded, will be posted to web site within 5 hours
- In-class questions welcome
 - Unmute and ask
 - Or raise hand (less clear to instructor!)
- HW will have on-line component, hand-in component and software component
 - On-line short questions with instant feedback using Blackboard quiz
 - SW will provide main program, input files, and output files for debugging.
You will develop needed functions.
- We will use Piazza for asynchronous question/answers at all hours
 - Enables all of you to benefit from answers

Course Organization - 4

- Course includes a project
 - Some topics will be suggested
 - Topics can be suggested by students
- Projects can be done in teams of up to 4 people
 - Or individually...Depends on how large a project one proposes
- Typical project: Develop a set of algorithms for solving a class of problems, implemented with different data structures, and evaluate the computation performance on realistic problem data.

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Course Organization - 5

- Policy on Cooperation
 - HW: Discussions with classmates, colleagues and googling on web are allowed in general
 - But you have to write down the solution yourself and fully understand what you write
 - Your code must be your own: base it on what you read on the web, but don't copy verbatim
 - Plagiarism tools will be used extensively. Any violations will result in course failure and referral to College Academic Council

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What is EC504 About?

- Data structures store data in manners that support efficient algorithm computation.
- Algorithms with appropriate data structures can improve problem solution efficiency by orders of magnitude.
- In this course, we study both advanced algorithms and appropriate data structures for solution of practical problems.
- We will focus on rigorous analysis, realistic applications and implementation.
- Software implementations of algorithms/data structures will be part of the course.

Course Outline

1. Review: Algorithm Analysis (CLRS 2-4)
 - Asymptotic complexity
 - Recursions
 2. Sorting: Classical and modern approaches (CLRS 4,9, Notes on TimSort)
 3. Review: Trees and their Properties (CLRS App. B)
 4. Efficient Search in one-dimension
 - Hash Tables (CLRS 11)
 - Balanced Search Trees (Red-Black, Splay, B-Trees, VEB trees) (CLRS 12,13,18,20)
 - Priority queues (Binary, Binomial, Fibonacci Heaps, VEB trees) (CLRS 6,19,20)
- std::sort()?*
Python?
Java? →
- Sgl: ?*
Oracle: ? ...

Course Outline - 2

6. Graphs and Network Optimization (CLRS 22)
 - Minimum spanning trees: Greedy algorithms (CLRS 23)
 - Shortest Paths: Greedy algorithms, Dynamic Programming (CLRS 24,25)
 - Max-flow: Ford-Fulkerson, Preflow-push (CLRS 26)
 - Min-cost flow: Assignment problems, auction algorithms, successive shortest path algorithms (Notes)
 - Applications (Notes)
7. Complexity Theory
 - NP-Complete Problems - Definition and examples (CLRS 34)
 - Approximation Algorithms - Knapsack, Traveling Salesperson, ... (CLRS 35)
8. Advanced Topics

What is an algorithm? An unambiguous list of steps (program) to transform some input into some output.



- Pick a Problem (set)
- Find method to solve
 1. Correct for all cases (elements of set) ✓
 2. Each step is finite ($\Delta t_{\text{step}} < \text{max time}$)
 - Next step is unambiguous
 - Terminate in finite number of steps ✓
- ◆ You know many examples:
GCD, Multiply 2 N bit integers, ...

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Bagdad (Iraq) 780-850

Growth of Algorithm Run Time with Size n : ↗ Big “Oh”

- $O(g(n))$: Set of functions of n that grow no faster than $g(n)$ as n gets large

$| g(n) \uparrow n \neq \infty$

→ $f(n) \in O(g(n))$ if and only if there exists $c > 0, N_0 > 0$
such that $|f(n)| \leq c|g(n)|$ whenever $n > N_0$

Alternative: $f(n) \in O(g(n)) \iff \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} < \infty$ ✓

$f(n)$ ∈ $O(g(n))$, or $f(n)$ is $O(g(n))$, or $f(n) \cancel{\in} O(g(n))$ ✓

$f(n) \in \cancel{O}(g(n)) \iff \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$

Growth of Algorithms: Other concepts

- $o(g(n))$: grow strictly slower than $g(n)$

$$f(n) \in o(g(n)) \iff \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$$

- $\Omega(g(n))$: grow no slower than $g(n)$ as n gets large

$f(n) \in \Omega(g(n))$ if and only if there exists $c > 0, N_0 > 0$ such that $|f(n)| \geq c|g(n)|$ whenever $n > N_0$

- $\omega(g(n))$: grow strictly faster than $g(n)$

$$f(n) \in \omega(g(n)) \iff \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$$

- $\Theta(g(n))$: grow similar to $g(n)$:

$$f(n) \in O(g(n)), f(n) \in \Omega(g(n))$$

Intuition



- $f = O(g)$
- $f = o(g)$
- $f = \Omega(g)$
- $f = \omega(g)$
- $f = \Theta(g)$
- $f \leq g$
- $f < g$
- $f \geq g$
- $f > g$
- $f = g$

$f \in O(g) \rightarrow$

$f \in \Theta(g) \rightarrow$

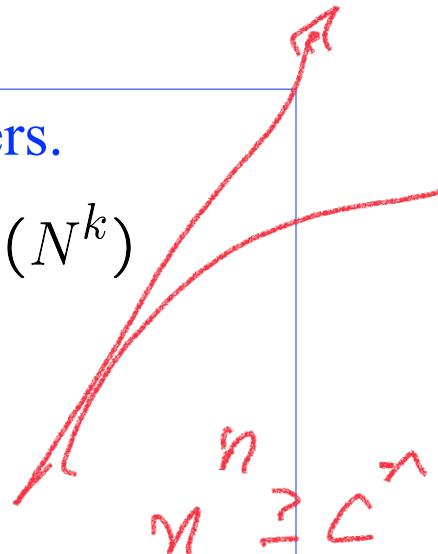
Rules of thumb

- For polynomials, only the largest term matters.

$$a_0 + a_1 N + a_2 N^2 + \dots + a_k N^k \in O(N^k)$$

- $\log N$ is in $o(N)$

Proof: As $N \rightarrow \infty$ the ratio $\log(N)/N \rightarrow 0$



- Some common functions in increasing order

1 $\log N$ \sqrt{N} N $N \log N$ N^2 N^3 N^{100} 2^N 3^N $N!$ N^N

- For $r \geq 1$, $d > 0$, $\Rightarrow n^d \in o(r^n)$

- For $b > 1$, $r > 0$, $\Rightarrow (\log(n))^b \in o(n^r)$

Some examples

- Which increases faster? What is the relation?

◻ ~~($100n^2$, $0.01 * 2^n$)~~

2^n .

◻ $(0.1 * \log n, 10n)$

2^n

◻ $(10^{10}n, 10^{-10}n^2)$

2^n

◻ $(0.01n^3 - 10, 100n^2)$

1^{st}

◻ $(\ln(300n^4), \ln(n))$

$\Theta(\cdot)$

$$\log(300n^4) = 4\log n + \underline{\log(300)}$$

◻ $(n^4 + 3n^2 + 1, n^{0.01})$

1^{st}

◻ $(2^n, 3^n)$

$\lim \frac{3^n}{2^n} = \lim \left(\frac{3}{2}\right)^n = \infty$

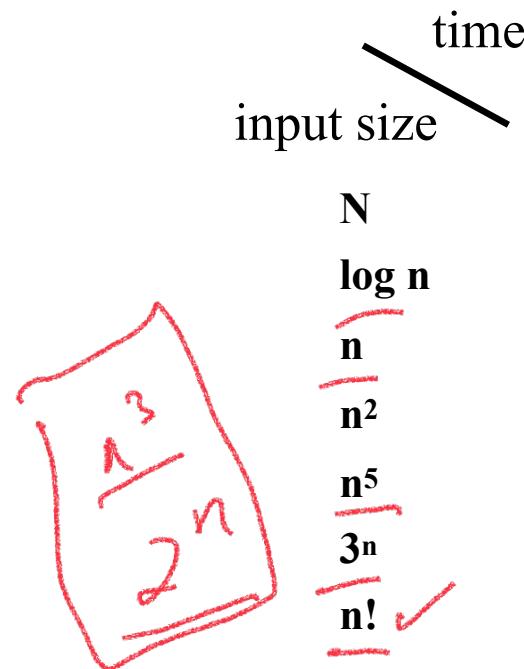
◻ (n^{10}, n^n)

2^{nd}

◻ $(n!, n^2)$ (Stirling's formula: $n! \approx (2\pi n)^{1/2} \left(\frac{n}{e}\right)^n$)

1^{st}

Why is big-O important?



input size	time	(processor doing ~1,000,000 steps per second)					
N	10	20	30	40	50	60	
$\log n$	3.3μsec	4.4μsec	5μsec	5.3μsec	5.6μsec	5.9μsec	
n	10μsec	20μsec	30μsec	40μsec	50μsec	60μsec	
n^2	100μsec	400μsec	900μsec	1.5msec	2.5msec	3.6msec	
n^5	0.1sec	3.2sec	24.3sec	1.7min	5.2min	13min	
3^n	59msec	48min	6.5yrs	385,500yrs	2x10 ⁸ centuries...		
$n!$	3sec	7.8x10 ⁸ millennia					

Non polynomial algorithms are terrible!
Logs are great! .

Analysis of Algorithms

- Algorithms don't have constant run time for a given input size
 - Run time depends on nature of input
 - Best-Case, Average Case, Worst Case
 - Typically care about worst case (other cases useful in practice)
- Example: Insertion Sort(a[0:n-1]): ✓

```
for (i=1; i < n; i ++){  
    key = a[i]; j = i-1;  
    while (j >= 0 && a[j] > key) {  
        a[j + 1] = a[j];  
        j = j - 1;  
    }  
    a[j + 1] = key;  
}
```

↓
O(n log n)

✓

Best? $O(n)$ ✓

Average $\Omega(n^2)$

Worst Case? $\Omega(n^2)$

$\sum_{i=1}^n i = \frac{n(n+1)}{2} \approx O(n^2)$

OUTER LOOP TRACE FOR INSERTION SORT:

	a[0]	a[1]	a[2]	a[3]	a[4]	a[5]	a[6]	a[7] (Swaps)
■	6	5	2	8	3	4	7	1 (1) ↗
	5 ←	→ 6						
■	5	6	2	8	3	4	7	1 (2)
	2 ←	→ 6						
	2 ←	→ 5						
■	2	5	6	8	3	4	7	1 (0)
■	2	5	6	8	3	4	7	1 (3)
■	2	3	5	6	8	4	7	1 (3)
■	2	3	4	5	6	8	7	1 (1)
■	2	3	4	5	6	7	8	1 (7) ↗
■	1	2	3	4	5	6	7	8 (17 total swaps)

Analysis of Algorithms

- Example: Insertion Sort(a[0:n-1]):

```
for (i=1; i < n; i++) {
    key = a[i]; j = i-1;
    while (j >= 0 && a[j] > key) {
        a[j + 1] = a[j];
        j = j - 1;
    }
    a[j + 1] = key;
}
```

- Analysis:

- Best case: input is sorted, so while loop ends after 2 ops: $\Theta(n)$
- Average case: while loop goes over 1/2 of list to find insert



- Worst case: input is sorted in reverse, so while look ends at $j = -1$.

```

for ( i = 0; i < N; i++) ↗
    for(j = i; j < N; j++) { →
        Sum = 0;
        for(k=i; k < j+1; k++) →
            Sum += a[k];
        if(Sum > MaxSum)
            MaxSum = Sum;
    }
}

```

$$\underbrace{\sum_{k=0}^n \sum_{j=0}^k}_{\text{double sum}} \sum_{i=0}^j a_i \rightarrow O(n^3)$$

```

for ( i = 0; i < N; i++) { ↗
    Sum = 0;
    for(j=i; j < N; j++) {
        Sum += a[j];
        if(Sum > MaxSum)
            MaxSum = Sum;
    }
}

```

$$O(n^2)$$

Different Algorithm

```
MergeSort(int *array, int l, int r):  
    if(l < r) {  
        int m = l+(r-l)/2;  
        mergeSort(array, l, m);  
        mergeSort(array, m+1, r);  
        merge(array, l, m, r);  
    }
```

$T(n)$

Recursive equation algorithm:

$T(n)$: run time for a list of size n . $m(n)$: time to merge arrays of length n

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

$m(n) \in \Theta(n)$

Merge Algorithm

```

void merge(int *array, int l, int m, int r) {
    int i, j, k, nl, nr;
    nl = m-l+1; nr = r-m;
    int larr[nl], rarr[nr];
    for(i = 0; i<nl; i++)
        larr[i] = array[l+i];
    for(j = 0; j<nr; j++)
        rarr[j] = array[m+1+j];
    i = 0; j = 0; k = l;
    //merge temp arrays to real array
    while(i < nl && j<nr) {
        if(larr[i] <= rarr[j]) {
            array[k] = larr[i];
            i++;
        } else{
            array[k] = rarr[j];
            j++;
        }
    }
    k++;
}

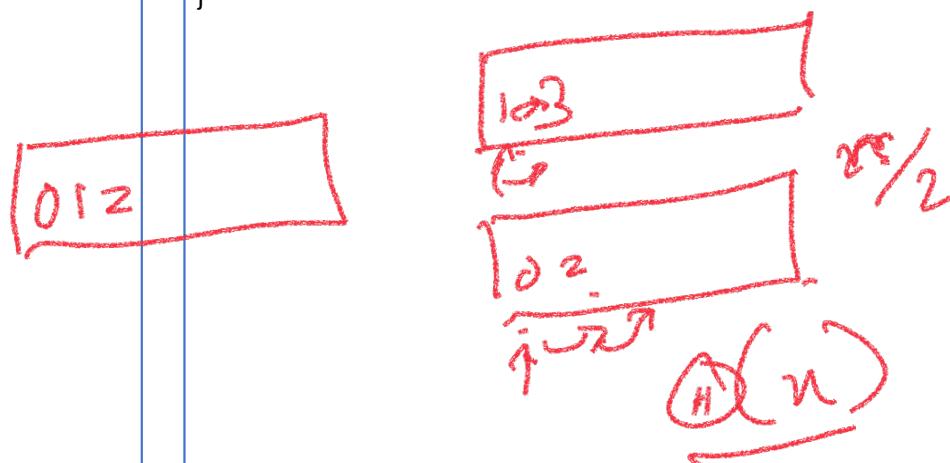
```

↙ ↘

```

while(i<nl) { //extra element in left array
    array[k] = larr[i];
    i++; k++;
}
while(j<nr) { //extra element in right array
    array[k] = rarr[j];
    j++; k++;
}

```



Mergesort Illustration

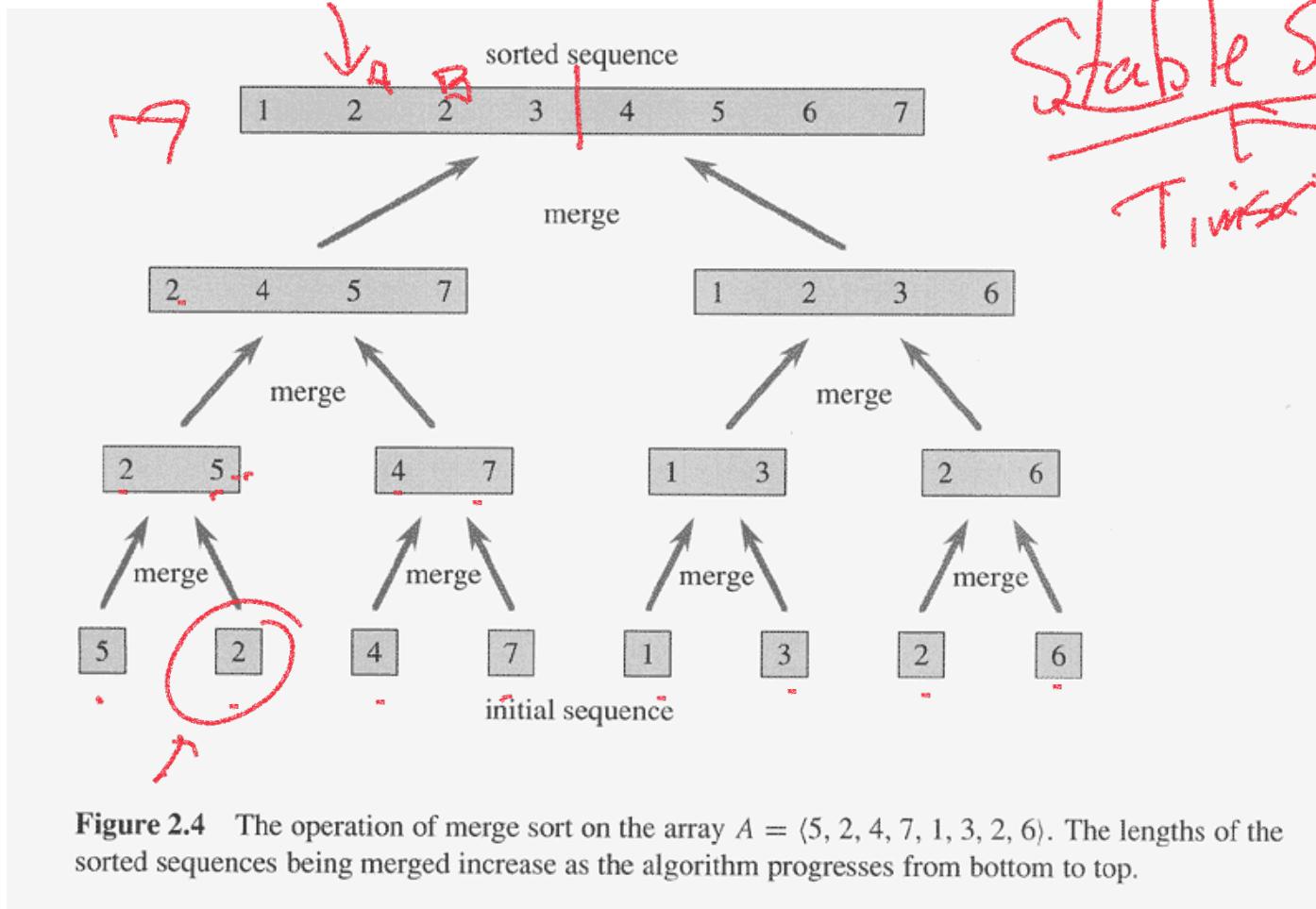


Figure 2.4 The operation of merge sort on the array $A = \{5, 2, 4, 7, 1, 3, 2, 6\}$. The lengths of the sorted sequences being merged increase as the algorithm progresses from bottom to top.

MASTER EQUATION: $T(n) = aT\left(\frac{n}{b}\right) + f(n)$

Theorem: The asymptotic Solution is:

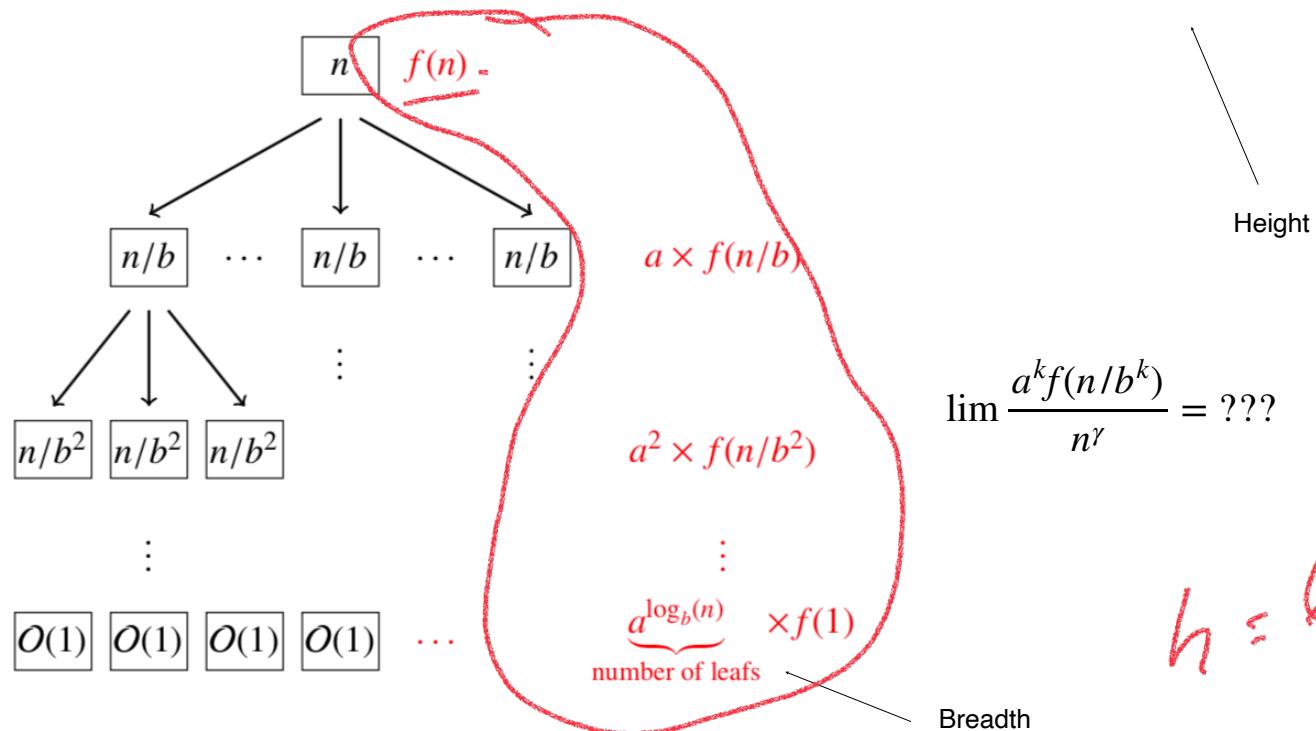
$$\begin{aligned} a &= 2 & \Rightarrow \gamma &= 1 \\ b &= 2 & x &= n \end{aligned}$$

- Define $\gamma = \log_b(a) \rightarrow a = b^\gamma$
 - Then, there are three cases
- - $T(n) \in \Theta(n^\gamma)$ if $f(n) \in O(n^{\gamma-\epsilon})$ for some $\epsilon > 0$
 - $T(n) \in \Theta(f(n))$ if $f(n) \in \Omega(n^{\gamma+\epsilon})$ for some $\epsilon > 0$
 - $T(n) \in \Theta(n^\gamma \log(n))$ if $f(n) \in \Theta(n^\gamma)$

BUILD TREE TO SOLVE

$$T(n) = aT(n/b) + f(n)$$

$$n/b^h = 1 \implies h = \log_b(n)$$



$$\rightarrow T(n) = f(n) + af(n/b) + \dots + a^{\log_b(n)-1}f(b^2) + a^h T(1)$$

LET'S ANALYZE FOR $n = b^h$, $f(n) \in O(n^k)$, so $f(n) = cn^k$

$$T(n) = f(n) + af(n/b) + \dots + a^{\log_b(n)-1}f(b^2) + a^h T(1) \quad \checkmark$$

$$T(n) = cn^k + ac(n/b)^k + a^2c(n/b^2)^k + \dots + a^{h-1}c(b^2)^k + a^h T(1) \quad \text{||.}$$

$$= cn^k + \left(\frac{a}{b^k}\right)cn^k + \left(\frac{a}{b^k}\right)^2 cn^k + \dots + \left(\frac{a}{b^k}\right)^{h-1} cn^k + a^h T(1).$$

Note: $a^h = n^\gamma$

$$cn^k + a^h = cn^k + n^\gamma = K n^\gamma + n^\gamma$$

If $k < \gamma$, then $b^k < a$. Case 1

If $k > \gamma$, then $b^k > a$. Case 2

If $k = \gamma$, then $b^k = a$. Case 3

$$n^\gamma \log(n).$$

Back to Mergesort

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

$$m(n) \in \Theta(n)$$

$T(n)$

Therefore, $a = 2, b = 2, \gamma = 1 \Rightarrow T(n) \in \Theta(n \log(n))$

Worst case for mergesort:

$$\Theta(n \log(n)) \xrightarrow{\text{arg. } \rightarrow O(n \log n)}$$

Average case for mergesort:

$$\Theta(n \log(n)) \xrightarrow{\text{avg. } \rightarrow O(n \log n)}$$

Best case for mergesort:

$$\Theta(n \log(n)) \xrightarrow{\text{best } \rightarrow O(n^2)}$$

worst: $O(n^2)$
best: $O(n \log n)$.

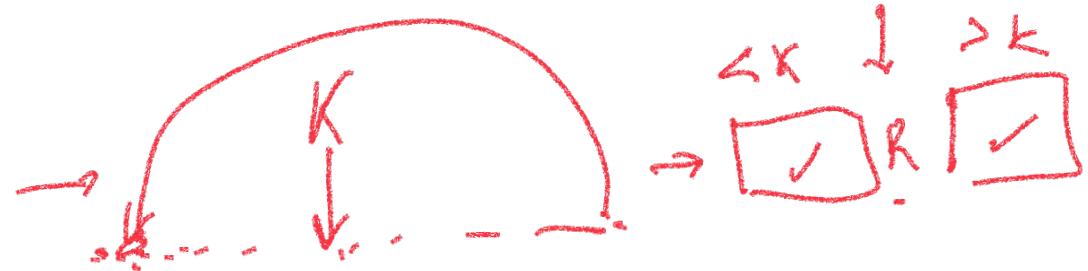
Order Statistics

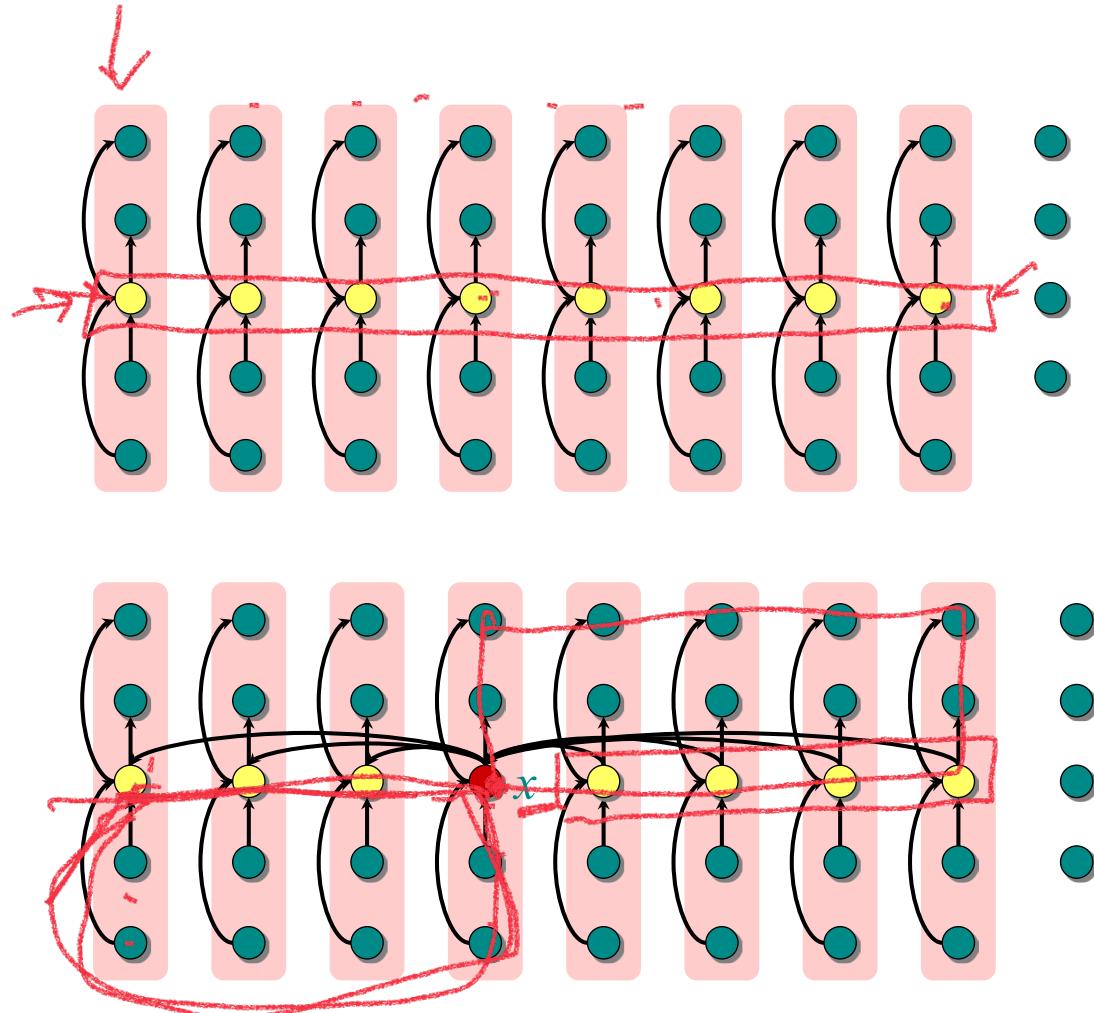
- Select the i-th smallest of n elements (the element with rank i).
 - $i = 1$: minimum;
 - $i = n$: maximum;
 - $i = \lfloor (n+1)/2 \rfloor$ or $\lceil (n+1)/2 \rceil$: median.
- Naive algorithm: Sort and index i th element.
- Worst-case running time = $\Theta(n \lg n) + \Theta(1) = \Theta(n \lg n)$,
 - using merge sort or Timsort or ...
- Can one do better?

$\Theta(n)$

Order Statistics

- $\text{SELECT}(i, n)$
 - Divide the n elements into groups of 5. Find the median of each 5-element group by rote.
 - Recursively SELECT the median x of the $\lfloor n/5 \rfloor$ group medians to be the pivot.
 - Partition around the pivot x . Let $k = \text{rank}(x)$.
 - if $i=k$ then return x else if $i < k$
 - recursively SELECT the i -th smallest element in the lower part
 - else recursively SELECT the $(i-k)$ th smallest element in the upper part





- At least half the group medians are $\leq x$, which is at least $\lfloor [n/5]/2 \rfloor = \lfloor n/10 \rfloor$ group medians. Therefore, at least $3 \lfloor n/10 \rfloor$ elements are $\leq x$.

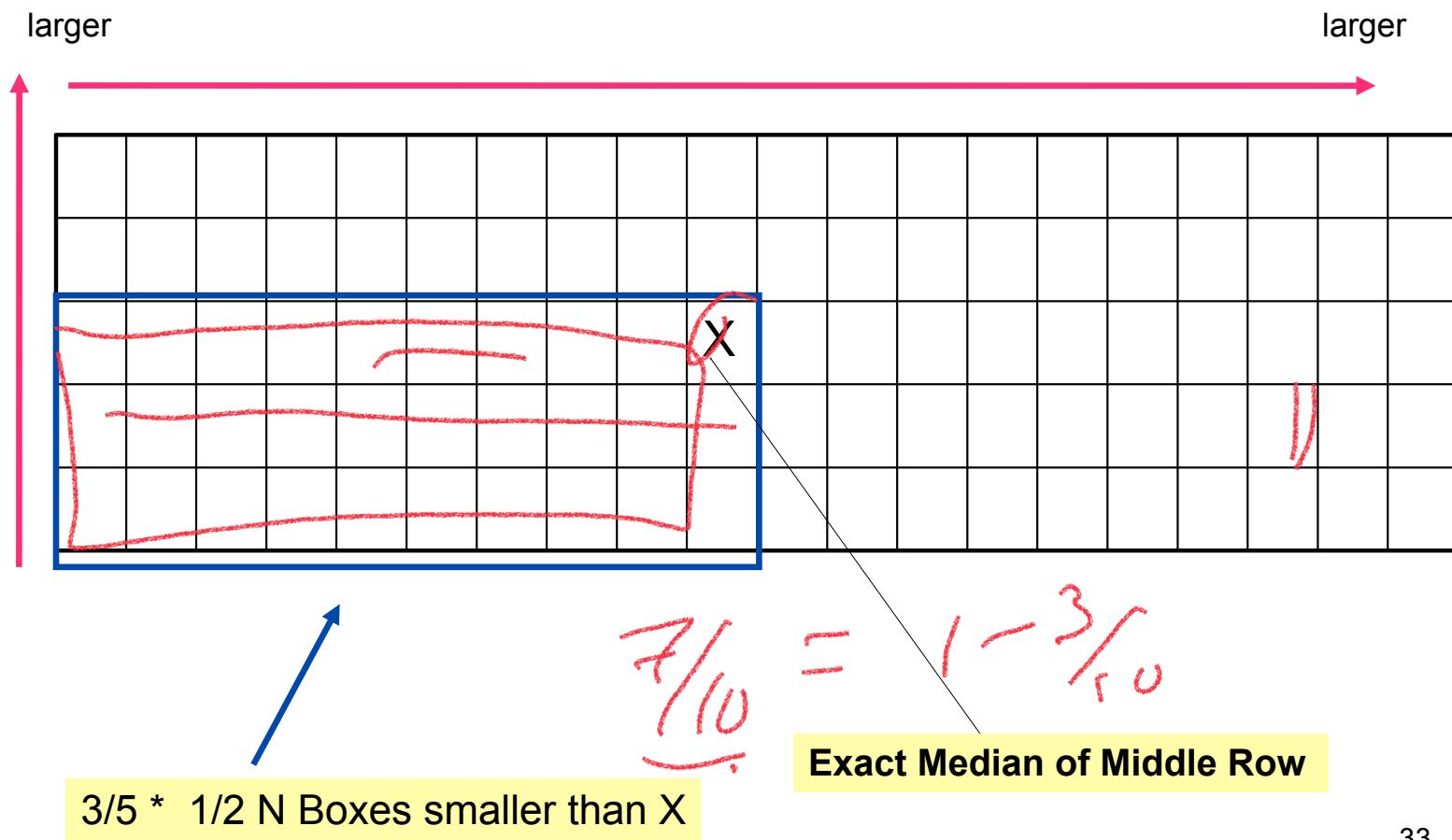
- Similarly, at least $3 \lfloor n/10 \rfloor$ elements are $\geq x$.

- Recursion:

$$T(n) \leq T(n/5) + T(7n/10) + cn$$

- Result: ~~$T(n) \in \Theta(n)$~~

5 row of N/5 Columns



Generalization of Master's Equation

Let $\alpha_1, \alpha_2, \dots, \alpha_k > 0$ be such that $\sum_{j=1}^k \alpha_j < 1$. Then,

If $\underline{T(n)} = \underline{T(\alpha_1 n)} + T(\alpha_2 n) + \dots + T(\alpha_k n) + \underline{cn}$, then $\underline{T(n)} \in \Theta(n)$

$$n^\delta < n$$

Other useful recursive solutions

Priority heaps...
Binomial or Fibonacci
Leonardo?...

→ $T(n) = a\underline{T(n - 1)} + b\underline{T(n - 2)}$ (Discrete difference equation, constant coefficient)

Can solve via z-transforms (if you know EC 401..) or follow method below:

Solution form: k^n .

Characteristic equation: Substitute into equation to get

$$k^n = ak^{n-1} + k^{n-2} \Rightarrow k^2 - ak - b = 0$$

Two solutions k_1, k_2 (or a repeated root).

If $k_1 \neq k_2$ then $T(n) = Ak_1^n + Bk_2^n$. Otherwise, $T(n) = Ak_1^n + Bnk_1^n$.

✓ te^{kt}

$n k_1^n$

Example: Fibonacci numbers

$$F(n) = F(n - 1) + F(n - 2) \quad \checkmark$$

Characteristic equation:

$$K^2 - K - 1 = 0$$

Solutions:

$$\frac{1}{2} \pm \frac{\sqrt{1+4}}{2} = \frac{1+\sqrt{5}}{2}, \frac{1-\sqrt{5}}{2}$$

Initial condition: $F(0) = 0; F(1) = 1$

$$\begin{aligned} A K_1^n + B K_2^n &= A + B = 0 \Rightarrow A = -B \\ A(\frac{1+\sqrt{5}}{2})^n - A(\frac{1-\sqrt{5}}{2})^n &= 1 \Rightarrow A = \frac{1}{\sqrt{5}} \end{aligned}$$

