CS180: Algorithms and Complexity

Professor: Raghu Meka (raghum@cs)

Plan for Today

Logistics

What is an algorithm?

Course goals

- "Algorithmic thinking"
 - Design and analysis

- Algorithmic lens: all areas of science
 - Applications

Core design principles and algorithms

Discussion sections

Discussion sections

Attend them!

Discussion sections

Attend them!

Solutions to homework and practice problems



I. Homework: 6 x 3



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2. Weekly quizzes: I0 x I



- I. Homework: 6 x 3
- 2. Weekly quizzes: I0 x I
- 3. Exam 1: 26



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- 2. Weekly quizzes: I0 x I
- 3. Exam 1: 26
- 4. Exam 2: 22



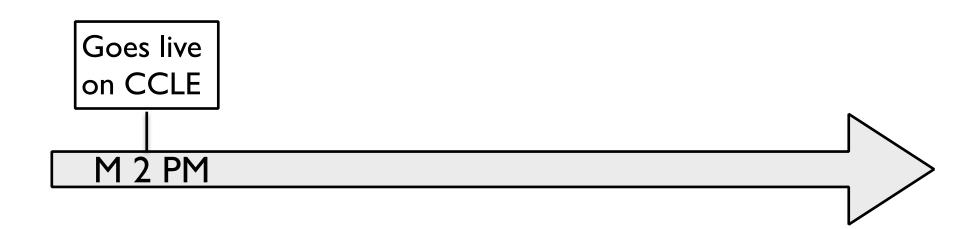
- I. Homework: 6 x 3
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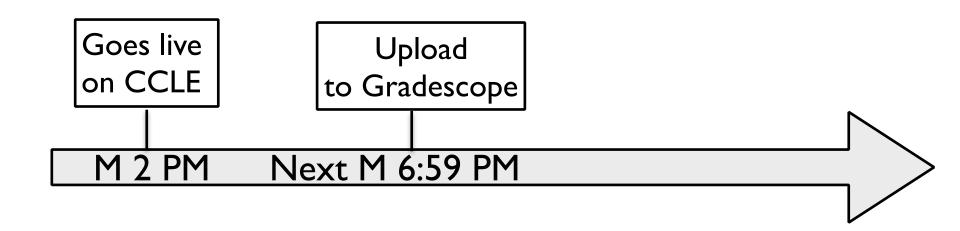


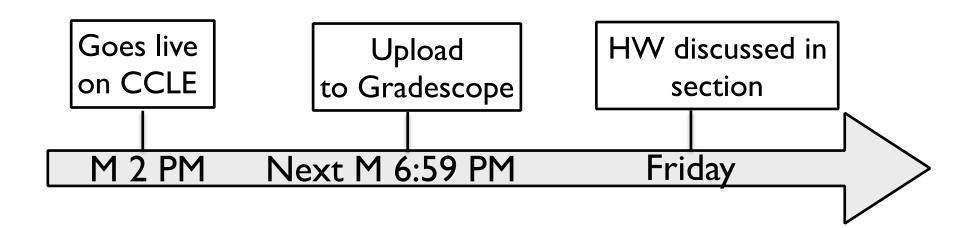
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- 4. Exam 2: 22
- 5. Exam 3: 24













I. Start early to use office hours



- 1. Start early to use office hours
- 2. Late-submissions = 0 credit



- 1. Start early to use office hours
- 2. Late-submissions = 0 credit
- 3. Regrade: Ask on Gradescope within a week



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4. Attempts count:

100% - correct; 75% - serious attempt;

50% - reasonable attempt.



I. Collaboration encouraged:

Must write your own answers.



- I. Collaboration encouraged:

 Must write your own answers.
- 2. Form study-groups

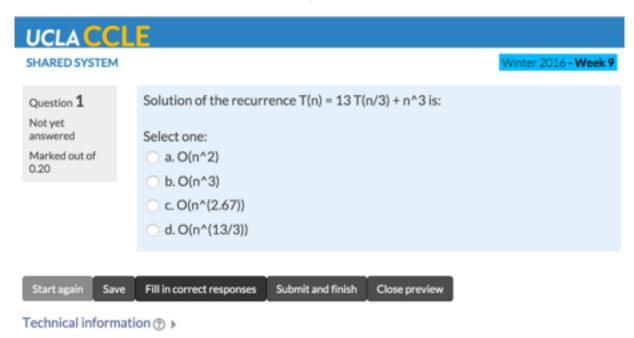


- I. Collaboration encouraged:

 Must write your own answers.
- 2. Form study-groups
- 3. Attempting homework honestly => Do well in exams.



Grading: Quizzes



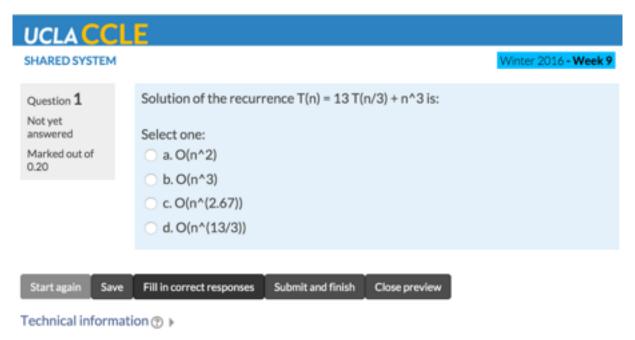
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Grading: Quizzes

UCLA CCL	<u> </u>	
SHARED SYSTEM		Winter 2016 - Week 9
Question 1 Not yet answered Marked out of 0.20	Solution of the recurrence T(n) = 13 T(n/3) + n^3 is: Select one: a. O(n^2) b. O(n^3) c. O(n^(2.67)) d. O(n^(13/3))	
Start again Save	Fill in correct responses Submit and finish Close preview	

1. Live on CCLE from W 6:00 PM to F 5:59PM

Grading: Quizzes



- I. Live on CCLE from W 6:00 PM to F 5:59PM
- 2. Less than five minutes; I point.

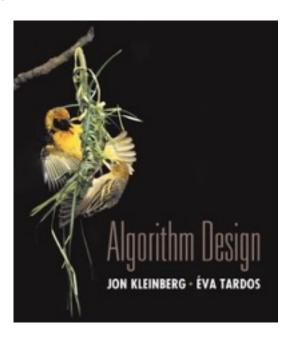
Lesson plan



Detailed Calendar: www.cs180.raghumeka.org

Required

Very good: read it!



Ask questions: Don't be convinced easily.



Ask questions: Don't be convinced easily. (especially about the WHY!)





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Office hours: high-latency, high-bandwidth

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Piazza: low-latency, low-bandwidth

Resources

Ask questions: Don't be convinced easily. (especially about the WHY!)

Office hours: high-latency, high-bandwidth

Piazza: low-latency, low-bandwidth

Tell us about how to make the course better

Plan for Today

Logistics

What is an algorithm?

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"A procedure for solving a mathematical problem (as of finding the greatest common divisor) in a finite number of steps that frequently involves repetition of an operation.



What is an algorithm?

"A procedure for solving a mathematical problem (as of finding the greatest common divisor) in a finite number of steps that frequently involves repetition of an operation.



"An algorithm is a finite, definite,effective procedure, with some inputand some output." — Donald Knuth





```
1 2 3 4 5 6
x 2 0 1 6
```

```
1 2 3 4 5 6
x 2 0 1 6
```

```
      I
      2
      3
      4
      5
      6

      x
      2
      0
      I
      6

      7
      4
      0
      7
      3
      6

      I
      2
      3
      4
      5
      6
      +

      0
      0
      0
      0
      0
      +
```

```
    I
    2
    3
    4
    5
    6

    x
    2
    0
    1
    6

    7
    4
    0
    7
    3
    6

    I
    2
    3
    4
    5
    6
    +

    0
    0
    0
    0
    0
    +
```

```
      I
      2
      3
      4
      5
      6

      x
      2
      0
      I
      6

      7
      4
      0
      7
      3
      6

      I
      2
      3
      4
      5
      6
      +

      0
      0
      0
      0
      0
      +
      -

      2
      4
      6
      9
      I
      2
      +
```

```
      1
      2
      3
      4
      5
      6

      x
      2
      0
      1
      6

      7
      4
      0
      7
      3
      6

      1
      2
      3
      4
      5
      6
      +

      0
      0
      0
      0
      0
      +
      -
      -

      2
      4
      6
      9
      1
      2
      +
      -
      6
```

```
      1
      2
      3
      4
      5
      6

      x
      2
      0
      1
      6

      7
      4
      0
      7
      3
      6

      1
      2
      3
      4
      5
      6
      +

      0
      0
      0
      0
      0
      +
      -
      2
      4
      6
      9
      1
      2
      +
      -
      -
      2
      9
      6
```

```
X 2 0 1 6

7 4 0 7 3 6

1 2 3 4 5 6 +

0 0 0 0 0 0 +

2 4 6 9 1 2 +

ANSWER = 2 4 8 8 8 7 2 9 6
```

First algorithm you ever saw/did ... Multiplication:

```
al a2 a3 a4 ... an x bl b2 b3 b4 ... bn
```

```
al a2 a3 a4 ... an x bl b2 b3 b4 ... bn
```

```
      a I a2 a3 a4 ...
      an

      x b I b2 b3 b4 ...
      bn
```

```
      a l a2 a3 a4 ...
      an

      x b l b2 b3 b4 ...
      bn

      c l c2 c3
      ... c<sub>n+1</sub>
```

```
      a l a2 a3 a4 ...
      an

      x b l b2 b3 b4 ...
      bn

      cl c2 c3
      ... c<sub>n+1</sub>

      +
```

```
      a l a2 a3 a4 ...
      an

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      c l c2 c3
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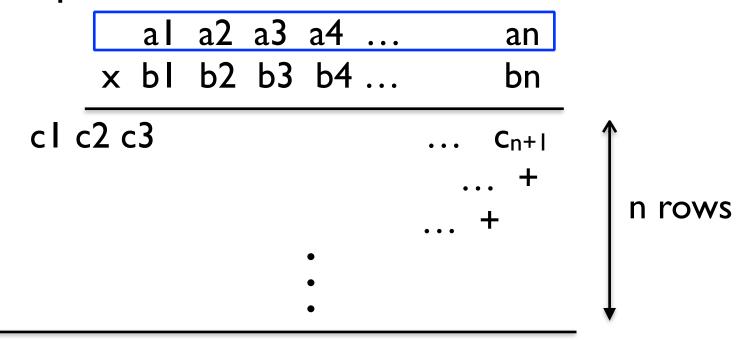
      +
```

```
      al a2 a3 a4 ...
      an

      x bl b2 b3 b4 ...
      bn

      cl c2 c3
      ... c<sub>n+1</sub>

      ... +
      ... +
```



```
a2 a3 a4 ...
                                     an
        x bl b2 b3 b4 ...
   cl c2 c3
                               ... C<sub>n+1</sub>
ANSWER =
```

Multiplication

INPUT: Two n-digit numbers a, b

OUTPUT: (a x b) in decimal format.

Multiplication

INPUT: Two n-digit numbers a, b

OUTPUT: $(a \times b)$ in decimal format.

How efficient is the previous algorithm?



What is efficiency?



What is efficiency?

Answer: Analysis of algorithms.



Runs fast on typical real-world inputs?

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PROS

Enough in practice

Easier to satisfy

Runs fast on typical real-world inputs?

PROS	CONS
Enough in practice	Hard to quantify
Easier to satisfy	Exceptions matter!
	"Real-world" is a moving target

Want a general theory of efficiency that is

I. Simple (mathematically)

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- 2. Objective (doesn't depend on i5 vs i7 processor)

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- 2. Objective (doesn't depend on i5 vs i7 processor)
- 3. Captures scalability (input-sizes change)
- 4. Predictive of practical performance
 - -"theoretically bad" algorithms should be bad in practice and vice versa (usually)

Most important resource in computing:

Most important resource in computing: TIME

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TIME: # of instructions executed in a simple programming language

• Only simple operations (+,*,-,=,if,call,...)

Most important resource in computing: TIME

- Only simple operations (+,*,-,=,if,call,...)
- Each operation takes one time step

Most important resource in computing: TIME

- Only simple operations (+,*,-,=,if,call,...)
- Each operation takes one time step
- Each memory access takes one time step

Most important resource in computing: TIME

- Only simple operations (+,*,-,=,if,call,...)
- Each operation takes one time step
- Each memory access takes one time step
- No fancy stuff built in (add these two matrices, copy this long string,...)

We left out things but...

Things we've dropped

```
memory hierarchy
```

disk, caches, registers have many orders of magnitude differences in access time

not all instructions take the same time in practice $(+, \div)$ communication

different computers have different primitive instructions

However,

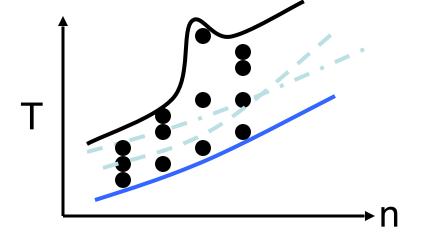
one can usually tune implementations so that the hierarchy, etc., is not a huge factor

Problem

 Algorithms can have different running times on different inputs.

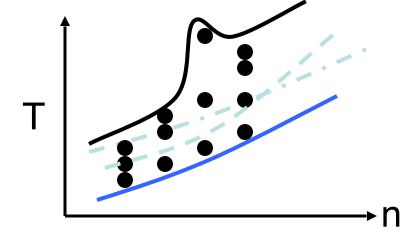
• Smaller inputs take less time, larger inputs take more time.

Solution



Measure performance on input size n

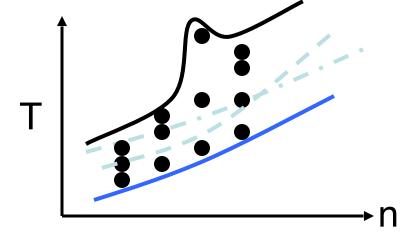
Solution



Measure performance on input size n

Average-case complexity: Average # steps algorithm takes on inputs of size n

Solution



Measure performance on input size n

Average-case complexity: Average # steps algorithm takes on inputs of size n

Worst-case complexity: Max # steps algorithm takes over all inputs of size n

Pros and cons:

Average-case

- Over what probability distribution? (different people may have different "average" problems)
- Analysis often hard

Worst-case

- + A fast algorithm has a comforting guarantee
- + Analysis easier than average-case
- + Useful in real-time applications (space shuttle, nuclear reactors)
- May be too pessimistic

Best-case ...

Best-case ...

Characterize growth-rate of (worst-case) run time as a function of problem size, up to a constant factor

Best-case ...

Characterize growth-rate of (worst-case) run time as a function of problem size, up to a constant factor

Why not try to be more precise?

Technological variations (computer, compiler, OS, ...)
 easily 10x or more

Complexity

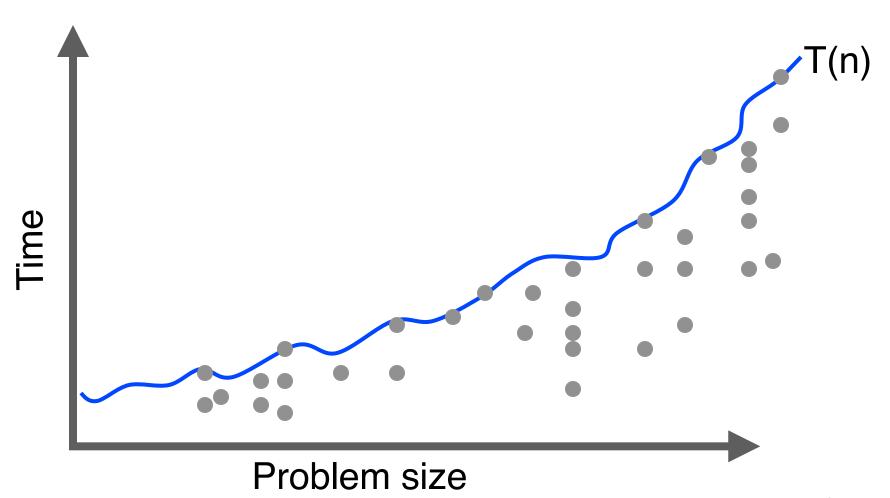
The *complexity* of an algorithm associates a number T(n) with each problem size n:

T(n) = worst-case time taken on problems of size n.

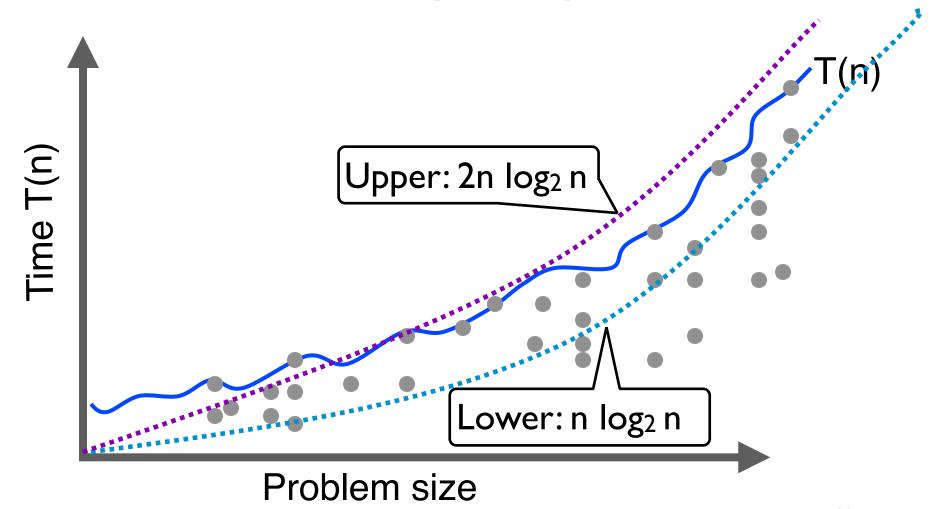
Mathematically: $T:\mathbb{N}\to\mathbb{R}^+$

i.e., T is a function that maps positive integers (problem sizes) to positive real numbers (number of steps).

Complexity



Complexity



Methodology for comparing run-times

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Given two functions $f, g : \mathbb{N} \to \mathbb{R}^+$

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```
f(n) = O(g(n)): iff there is a constant c>0 so that f(n) is eventually-always at most c g(n)
```

Methodology for comparing run-times Given two functions $f,g:\mathbb{N} \to \mathbb{R}^+$

f(n) = O(g(n)): iff there is a constant c>0 so that f(n) is eventually-always at most c g(n)

 $f(n) = \Omega(g(n))$: iff there is a constant c>0 so that f(n) is eventually-always at least c g(n)

Methodology for comparing run-times Given two functions $f,g:\mathbb{N}\to\mathbb{R}^+$

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 $f(n) = \Omega(g(n))$: iff there is a constant c>0 so that f(n) is eventually-always at least c g(n)

 $f(n) = \Theta(g(n))$: iff both hold - there are constants c_1 , $c_2 > 0$ so that eventually always $c_1g(n) < f(n) < c_2g(n)$

Examples

$$10n^2$$
-16n+100 is $O(n^2)$
 $10n^2$ -16n+100 < $10n^2$ for all n > 10

$$10n^2$$
-16n+100 is $\Omega(n^2)$

$$10n^2$$
-16n+100 > n^2 for all n > 10
Therefore also $10n^2$ -16n+100 is $\Theta(n^2)$

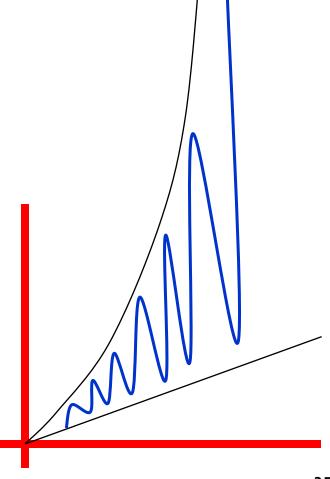
 $10n^2$ -16n+100 is not O(n) also not $\Omega(n^3)$

Big-Theta, etc. not always "nice"

$$f(n) = \begin{cases} n^2, & n \text{ even} \\ n, & n \text{ odd} \end{cases}$$

f(n) is not $\Theta(n^a)$ for any a.

Fortunately, such nasty cases are rare



Polynomial time

P: Running time is $O(n^d)$ for some constant d independent of the input size n.

NICE SCALING: Doubling problem-size, increases time by a constant factor c (e.g., $c \sim 2^d$)

Contrast with exponential: Doubling problem-size can square the run-time!

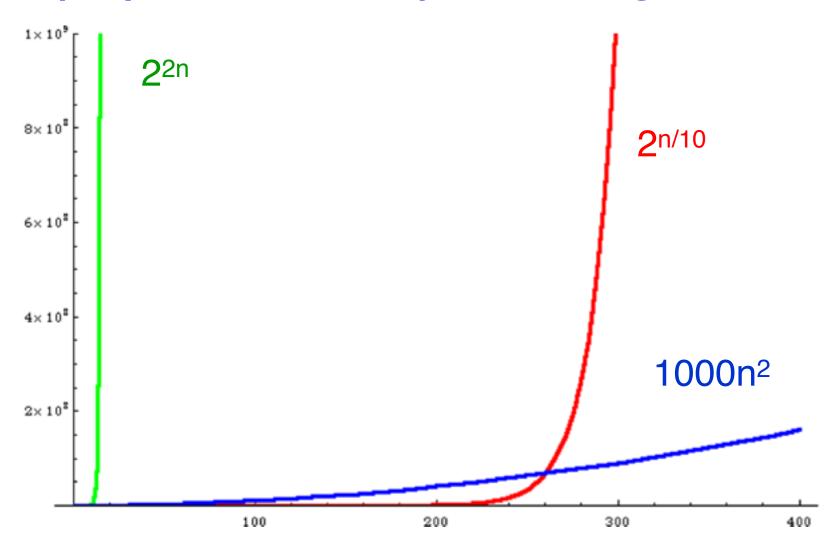
(e.g.,
$$T(n) = 2^{n/10}$$
 vs $T(2n) = 2^{2(n/10)} = T(n)^2$)

Polynomial time

P: Running time is $O(n^d)$ for some constant d independent of the input size n.

Behaves well under composition: if algorithm has a polynomial running time with polynomial number of calls to a subroutine with polynomial running time, then overall running time is still polynomial.

polynomial vs exponential growth



Why It Matters

Table 2.1 The running times (rounded up) of different algorithms on inputs of increasing size, for a processor performing a million high-level instructions per second. In cases where the running time exceeds 10²⁵ years, we simply record the algorithm as taking a very long time.

	n	$n \log_2 n$	n^2	n^3	1.5 ⁿ	2^n	n!
n = 10	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	4 sec
n = 30	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	18 min	10^{25} years
n = 50	< 1 sec	< 1 sec	< 1 sec	< 1 sec	11 min	36 years	very long
n = 100	< 1 sec	< 1 sec	< 1 sec	1 sec	12,892 years	10^{17} years	very long
n = 1,000	< 1 sec	< 1 sec	1 sec	18 min	very long	very long	very long
n = 10,000	< 1 sec	< 1 sec	2 min	12 days	very long	very long	very long
n = 100,000	< 1 sec	2 sec	3 hours	32 years	very long	very long	very long
n = 1,000,000	1 sec	20 sec	12 days	31,710 years	very long	very long	very long

not only get very big, but do so *abruptly*, which likely yields erratic performance on small instances 39

another view of poly vs exp

Next year's computer will be 2x faster. If I can solve problem of size n_0 today, how large a problem can I solve in the same time next year?

Complexity	Increase	E.g. $T=10^{12}$
O(n)	$n_0 \rightarrow 2n_0$	$10^{12} \rightarrow 2 \times 10^{12}$
$O(n^2)$	$n_0 \rightarrow \sqrt{2} n_0$	$10^6 \rightarrow 1.4 \times 10^6$
$O(n^3)$	$n_0 \rightarrow \sqrt[3]{2} n_0$	$10^4 \rightarrow 1.25 \times 10^4$
$2^{n/10}$	$n_0 \rightarrow n_0 + 10$	400 → 410
2 ⁿ	$n_0 \rightarrow n_0 + 1$	40 → 41

Summary

Typical initial goal for algorithm analysis is to find a

reasonably tight

asymptotic

bound on

worst case running time
as a function of problem size

i.e., Θ if possible

i.e., Θ or Θ

usually upper bound

This is rarely the last word, but often helps separate good algorithms from blatantly bad ones - so you can concentrate on the good ones!

Complexity of multiplication

Multiplication

INPUT: Two n-digit numbers a, b

OUTPUT: (a x b) in decimal format.

Complexity of multiplication

Multiplication

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OUTPUT: $(a \times b)$ in decimal format.

How efficient is grade-school algorithm?

First algorithm you ever saw/did ... Multiplication:

```
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```

```
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```
      a I a2 a3 a4 ...
      an

      x b I b2 b3 b4 ...
      bn
```

```
      a l a2 a3 a4 ...
      an

      x b l b2 b3 b4 ...
      bn

      c l c2 c3
      ... c<sub>n+1</sub>
```

```
      a l a2 a3 a4 ...
      an

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```

```
      a l a2 a3 a4 ...
      an

      x b l b2 b3 b4 ...
      bn

      cl c2 c3
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```

```
      a l a2 a3 a4 ...
      an

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      bn

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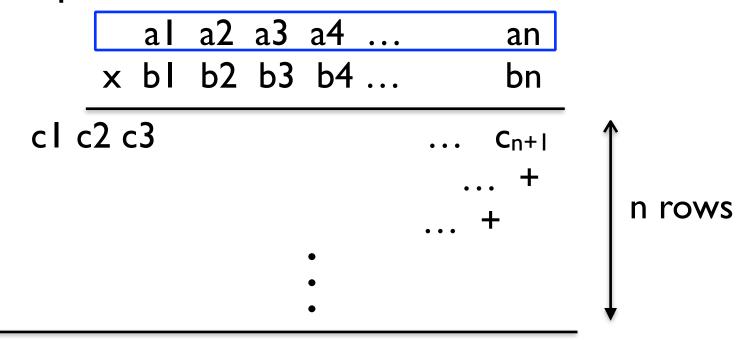
      +
```

```
      al a2 a3 a4 ...
      an

      x bl b2 b3 b4 ...
      bn

      cl c2 c3
      ... c<sub>n+1</sub>

      ... +
      ... +
```



```
a2 a3 a4 ...
                                     an
        x bl b2 b3 b4 ...
   cl c2 c3
                               ... C<sub>n+1</sub>
ANSWER =
```

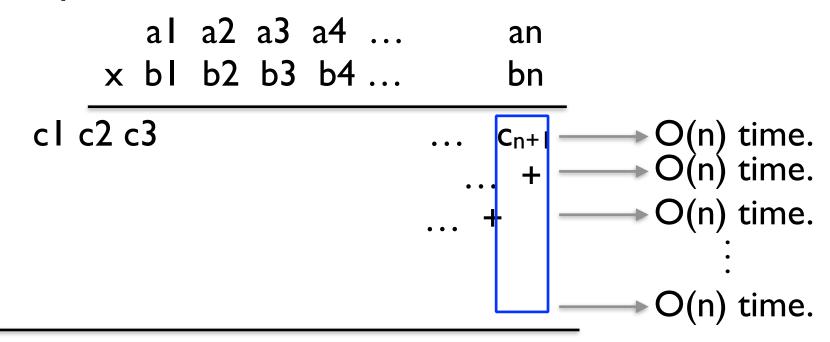
```
      al a2 a3 a4 ...
      an

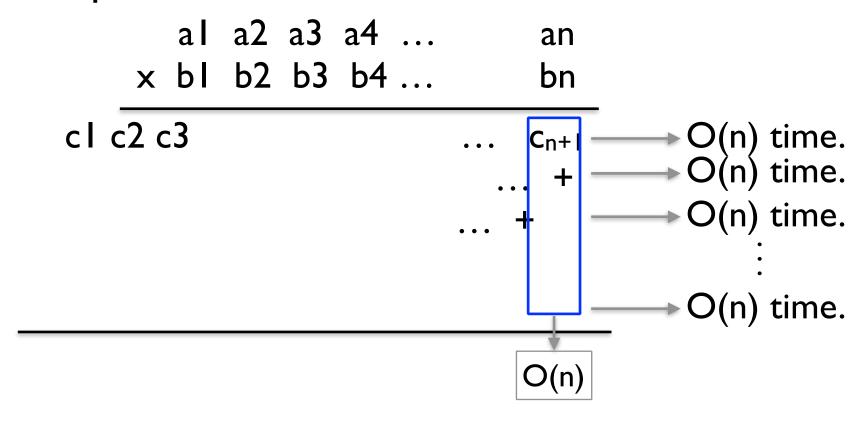
      x bl b2 b3 b4 ...
      bn

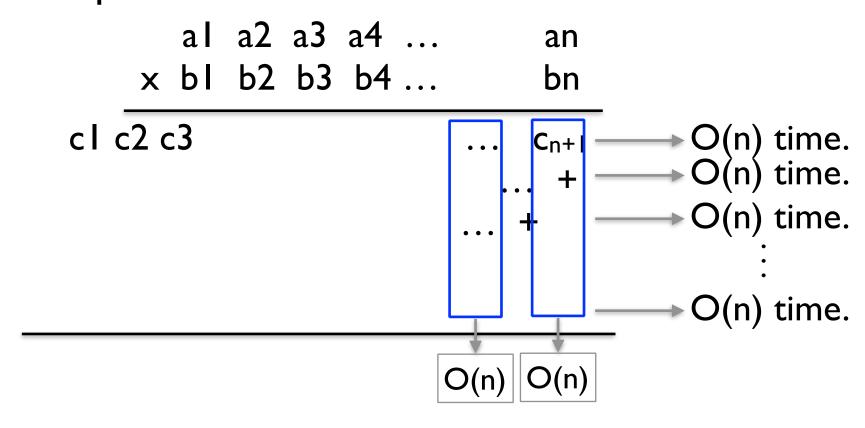
      cl c2 c3
      ... c<sub>n+1</sub>

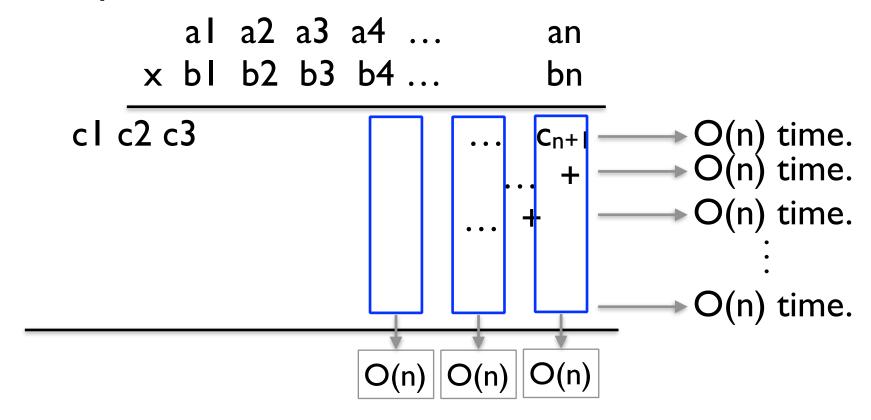
      ... +
      ... +

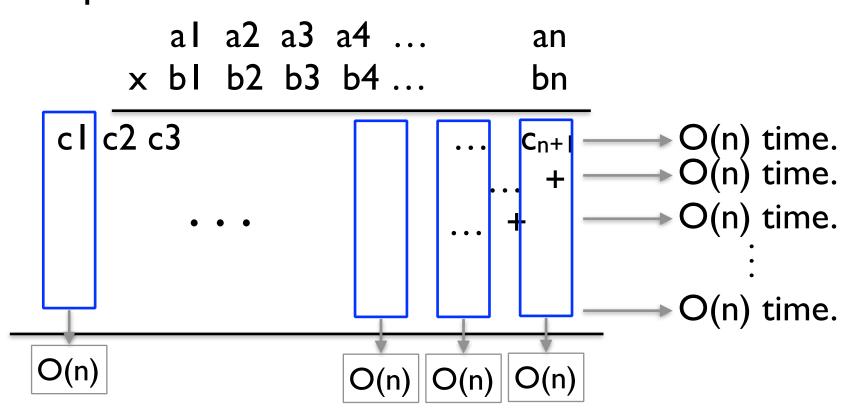
      ... +
      ... +
```

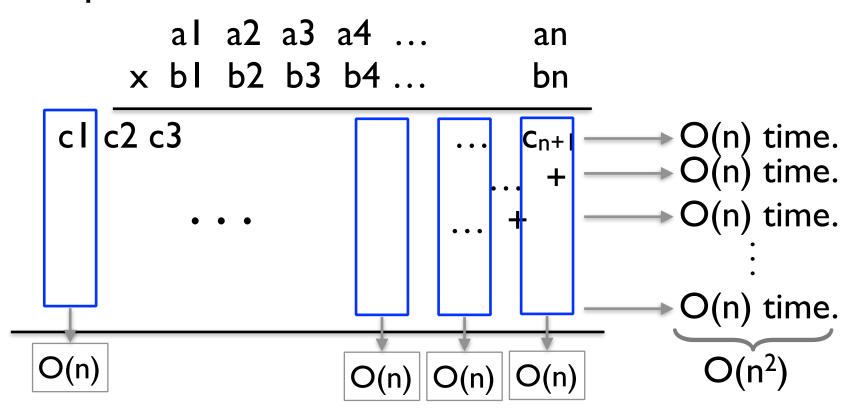


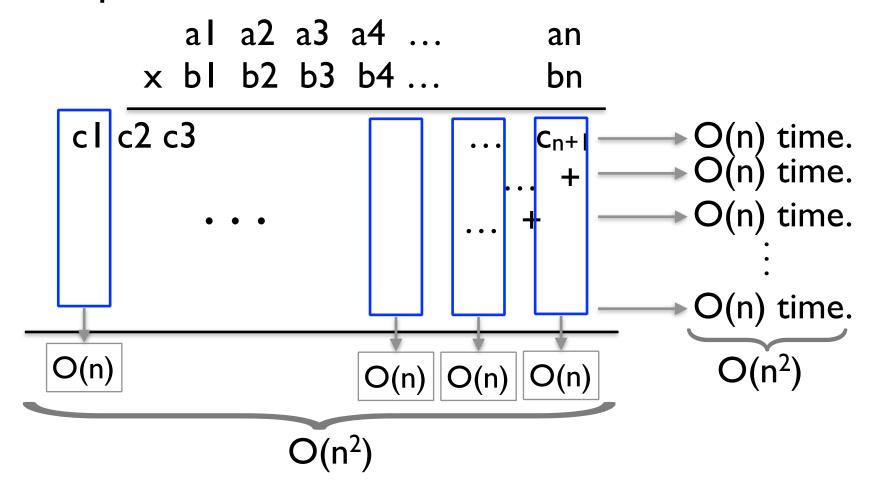


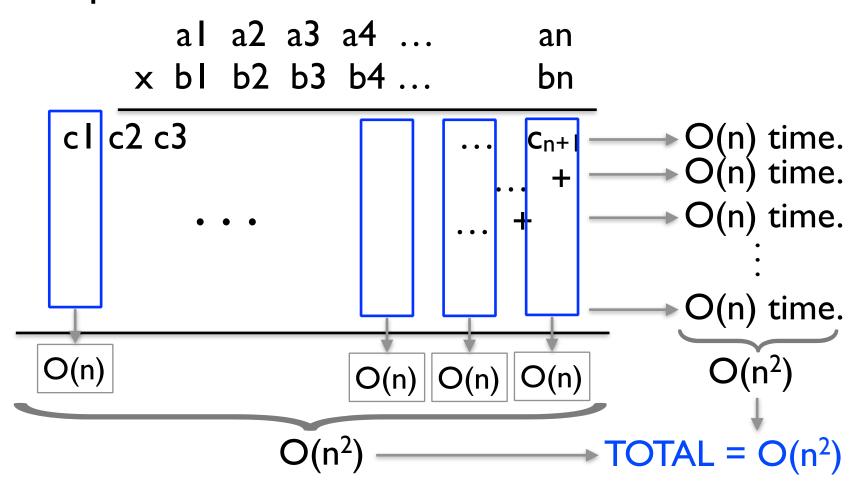












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Multiplication

INPUT: Two n-digit numbers a, b

OUTPUT: (a x b) in decimal format.

Complexity of multiplication

Multiplication

INPUT: Two n-digit numbers a, b

OUTPUT: $(a \times b)$ in decimal format.

Grade-school algorithm takes O(n²) time.

Divide-and-conquer paradigm

Divide-and-conquer.

- Divide problem into several subproblems.
- Solve each subproblem recursively.
- Combine solutions to subproblems into overall solution.

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- Divide problem into several subproblems.
- Solve each subproblem recursively.
- Combine solutions to subproblems into overall solution.

Most common usage.

- Divide size *n* into two subproblems of size n/2 in linear time.
- Solve two subproblems recursively.
- Combine two solutions into overall solution in linear time.

$$RUN-TIME = O(n log n)!$$

Sorting problem

Problem. Given a list of n elements from a totally-ordered universe, rearrange them in ascending order.



Sorting applications

Obvious applications.

- Organize an MP3 library.
- Display Google PageRank results.
- List RSS news items in reverse chronological order.

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Obvious applications.

- Organize an MP3 library.
- Display Google PageRank results.
- List RSS news items in reverse chronological order.

Some problems become easier once elements are sorted.

- Identify statistical outliers.
- Binary search in a database.
- Remove duplicates in a mailing list.

Sorting applications

Non-obvious applications.

- Convex hull.
- Closest pair of points.
- Interval scheduling / interval partitioning.
- Minimum spanning trees (Kruskal's algorithm).
- Scheduling

Sorting

input

A L G O R I T H M S										
	Α	L	G	0	R	ı	Т	Н	M	S

Sorting

• INPUT: Array

input



Sorting

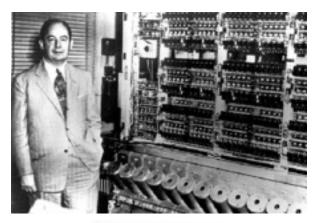
- INPUT: Array
- OUTPUT: Sorted array

input



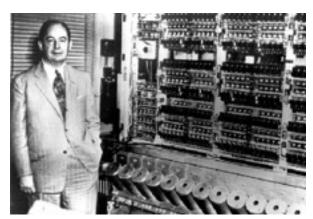
output

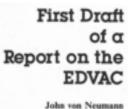




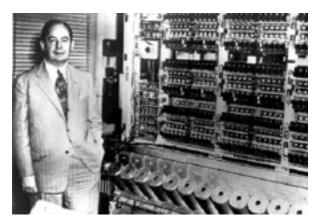
First Draft of a Report on the EDVAC

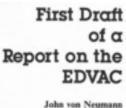
Developed in 1945 by von Neumann



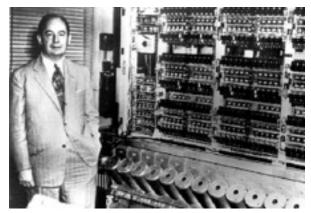


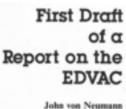
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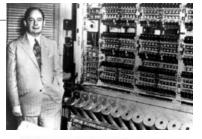
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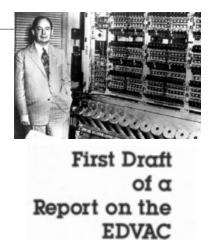
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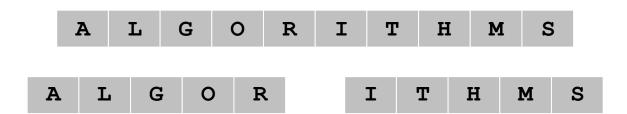
Excellent example of principles of Divide & Conquer



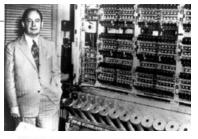
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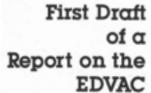
Divide array into two halves.

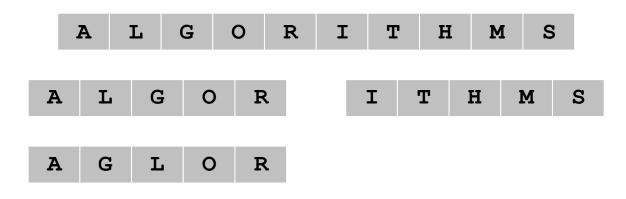




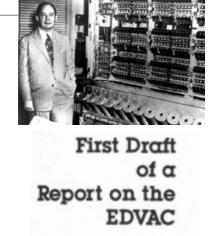
- Divide array into two halves.
- Recursively sort left half.

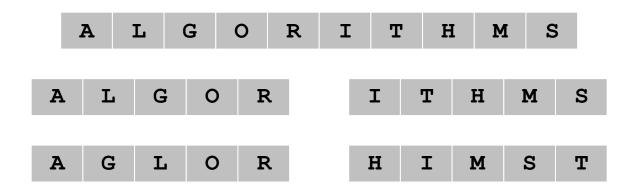




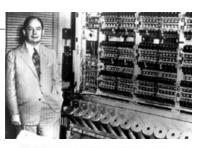


- Divide array into two halves.
- Recursively sort left half.
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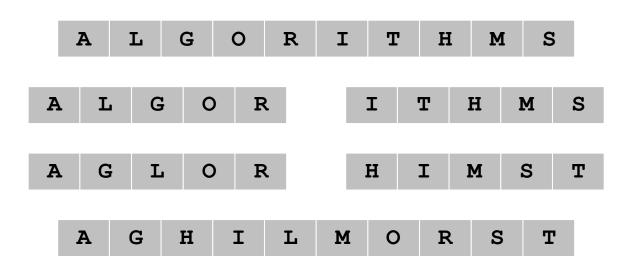




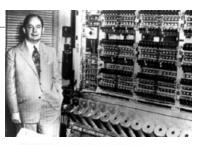
- Divide array into two halves.
- Recursively sort left half.
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- Merge two halves to make sorted whole.



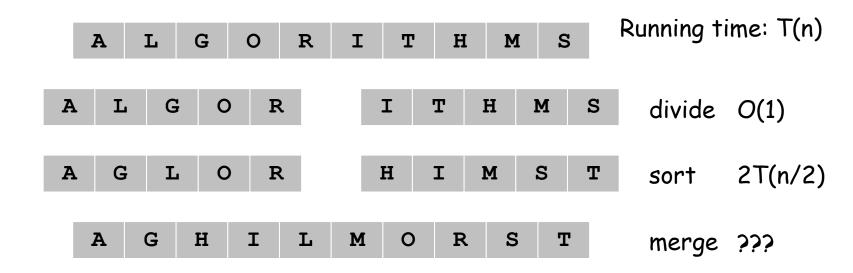
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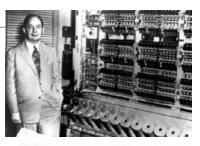
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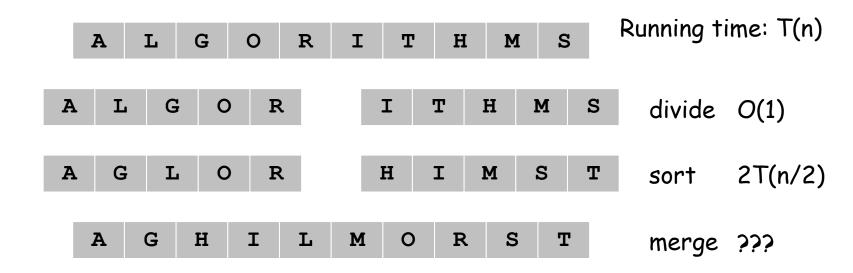
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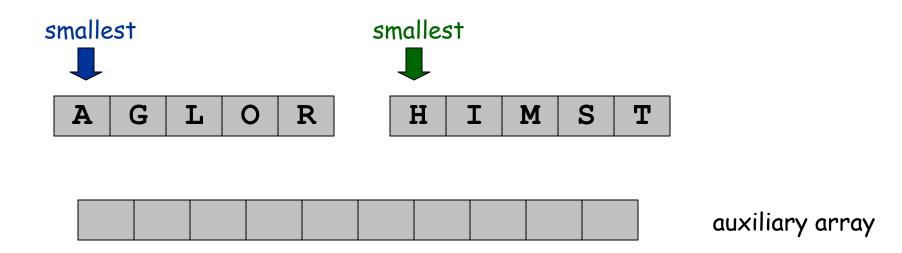


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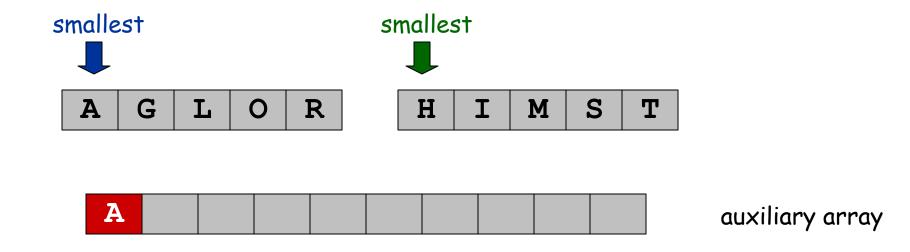


- Keep track of smallest element in sorted halves
- Insert smallest of two elements into new array
- Repeat until done

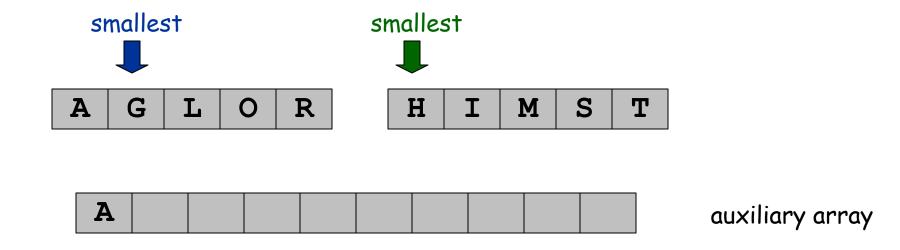
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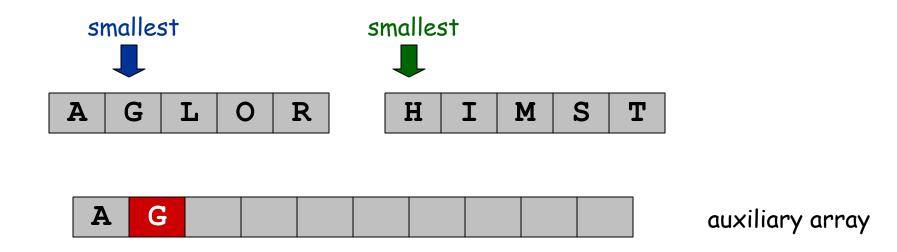
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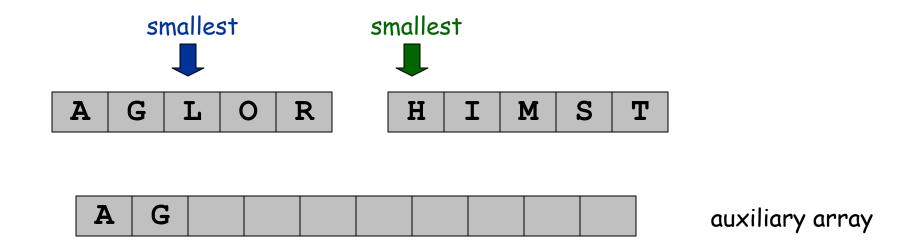
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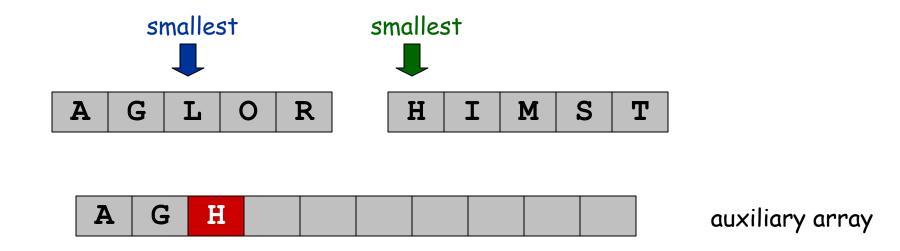
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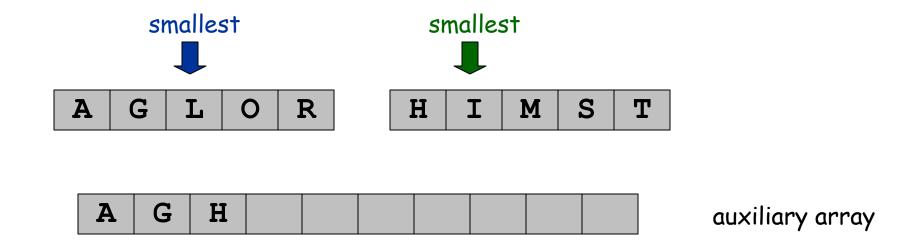
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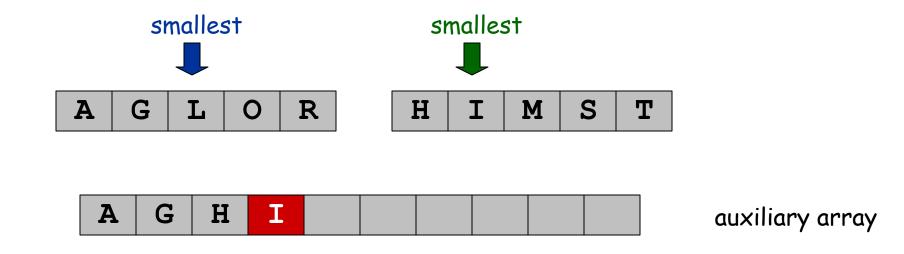
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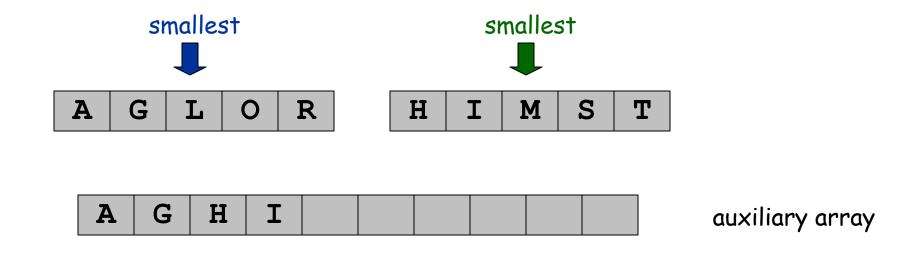
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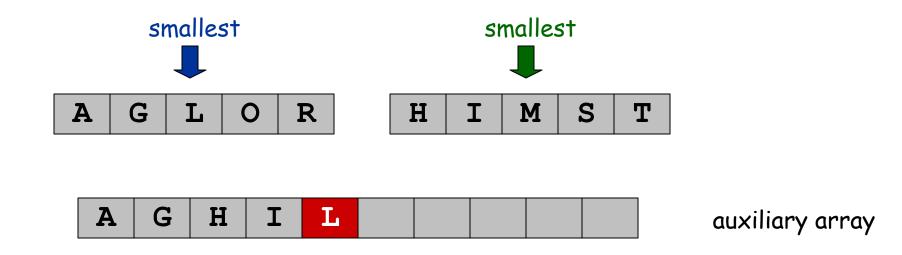
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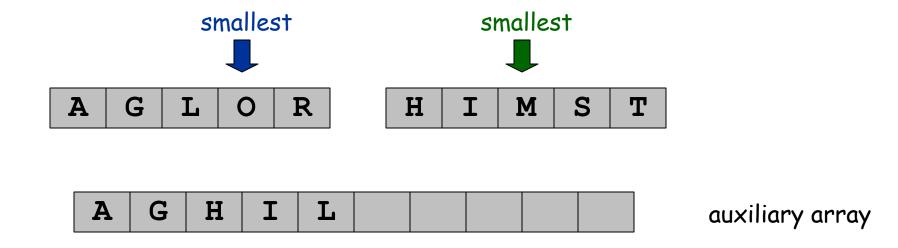
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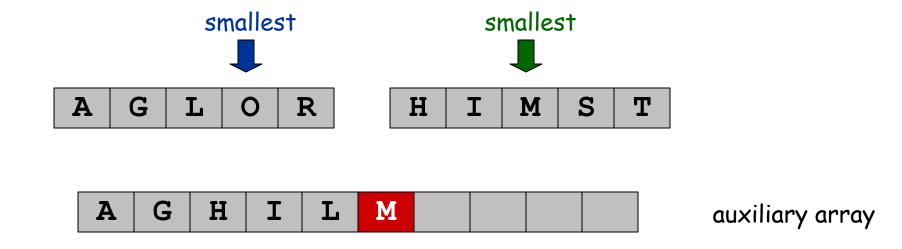
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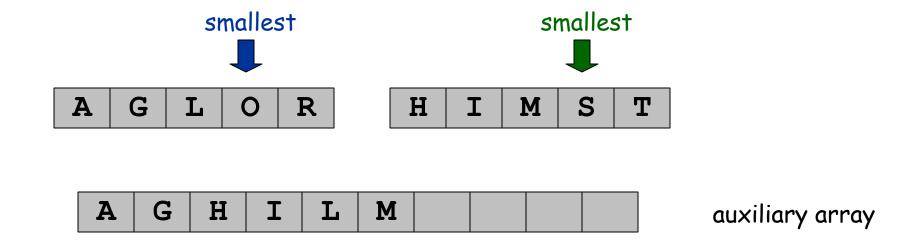
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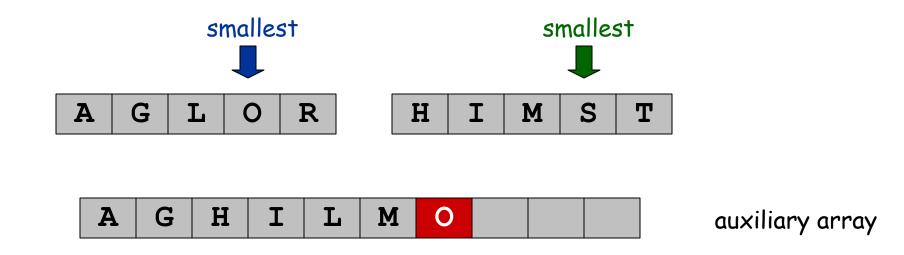
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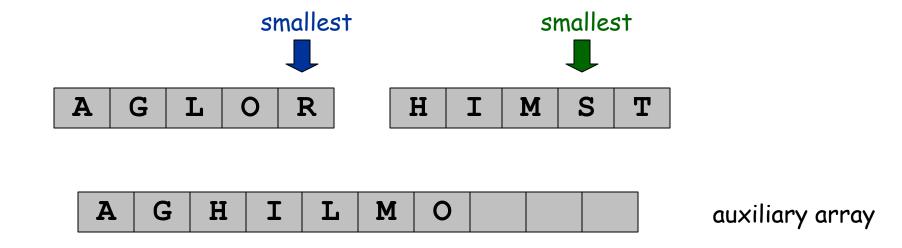
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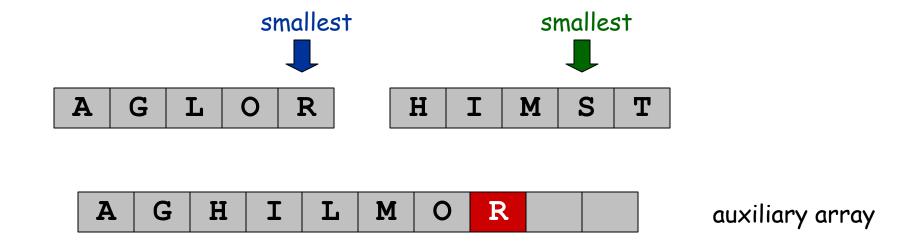
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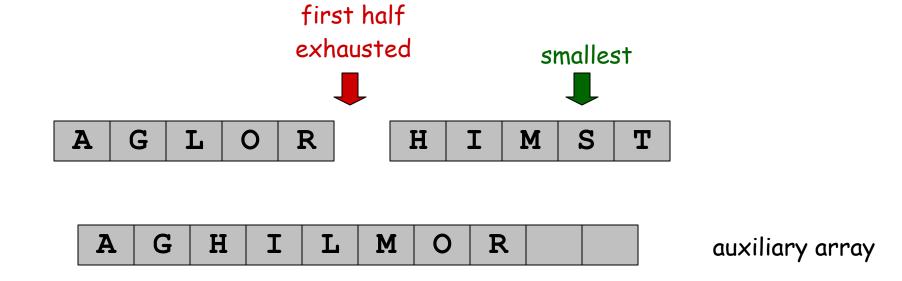
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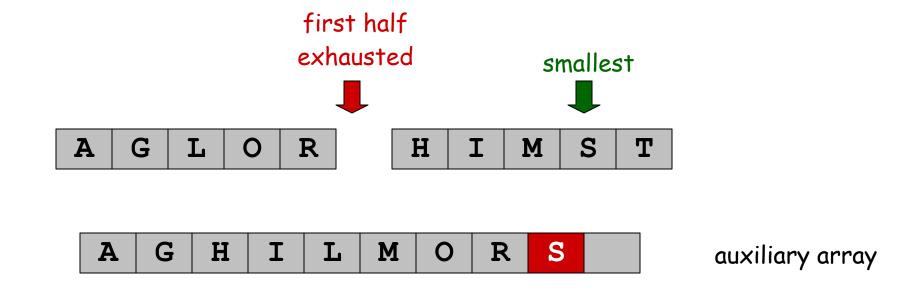
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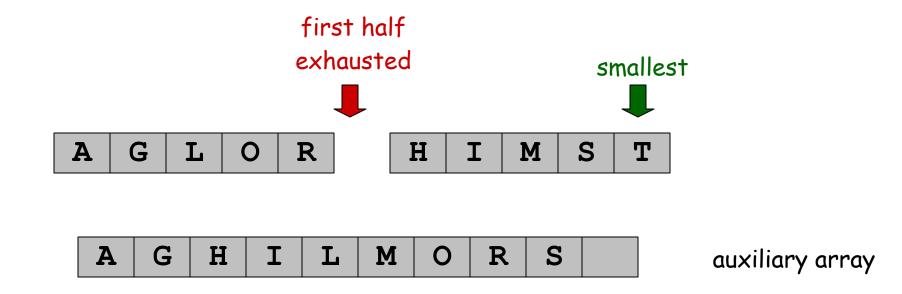
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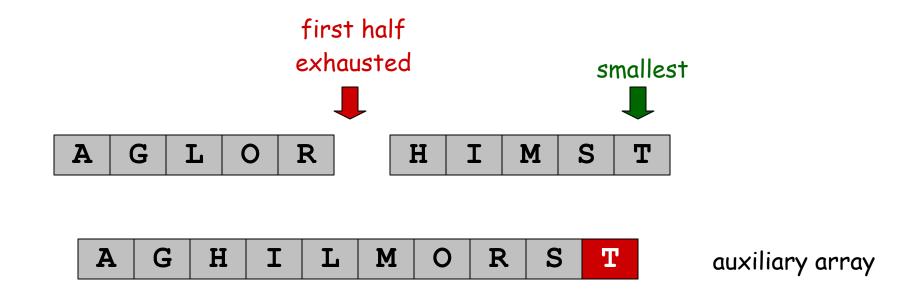
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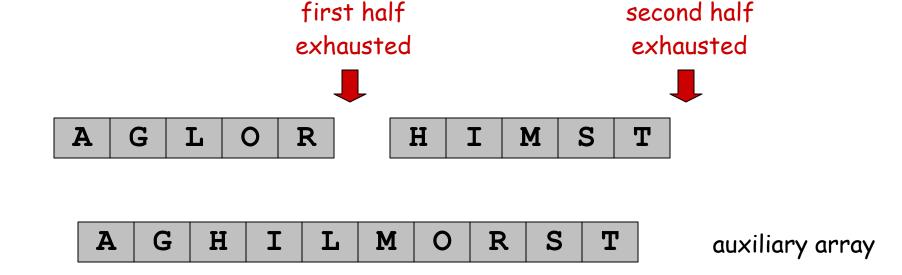
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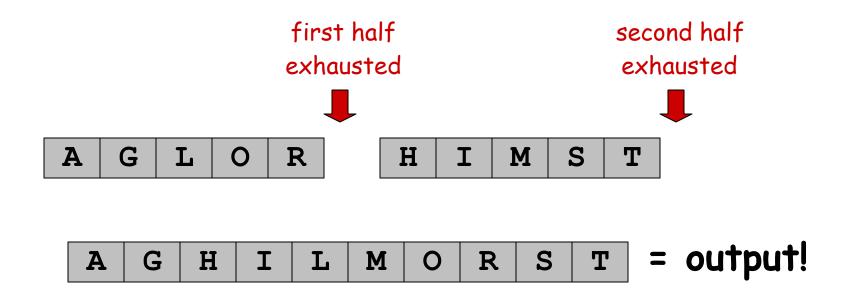
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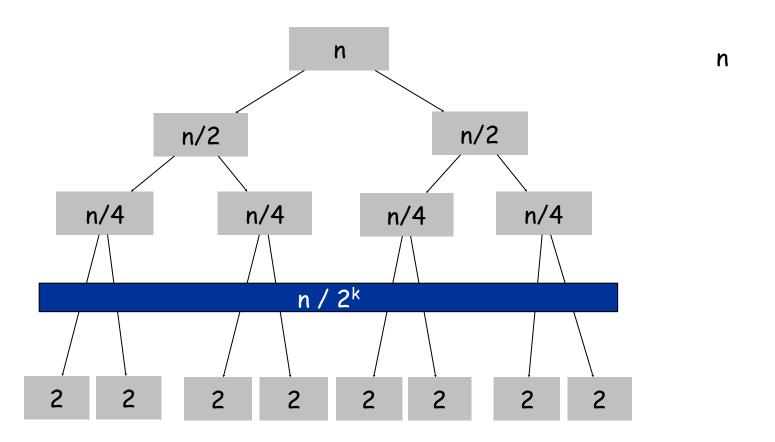
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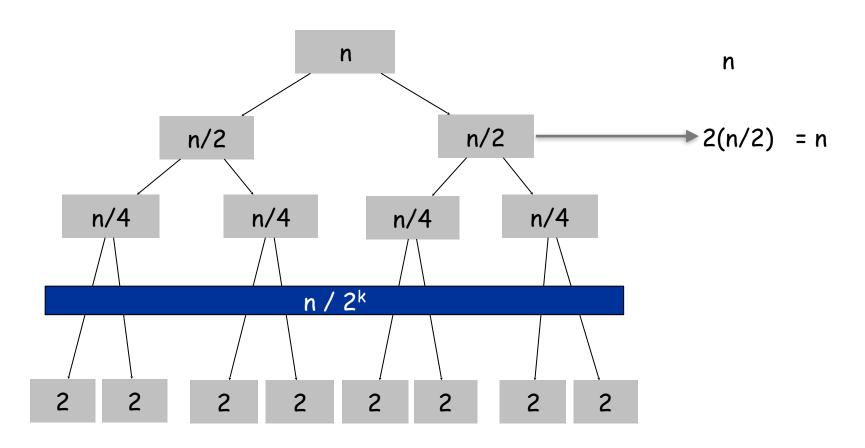
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Solution: O(n log n)

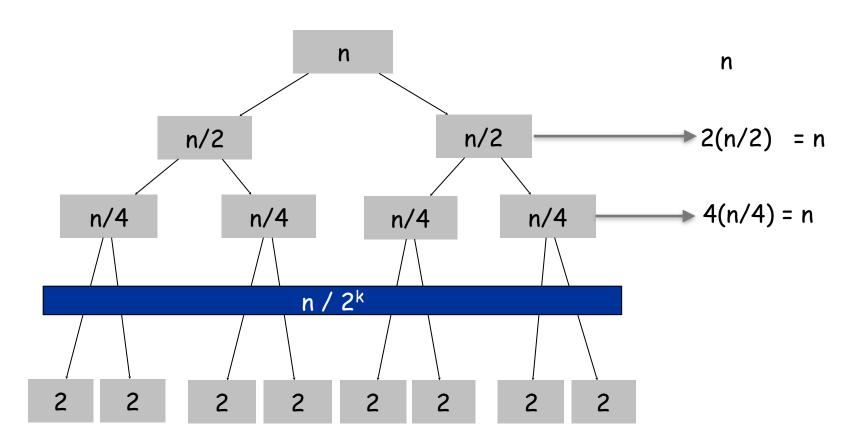
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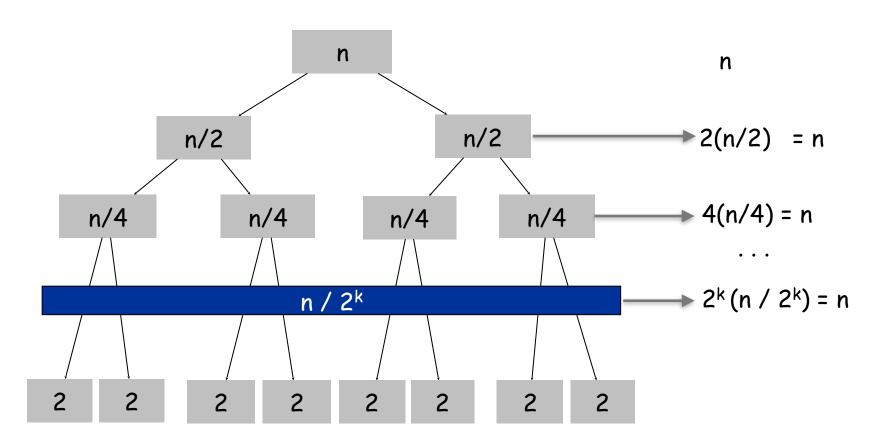
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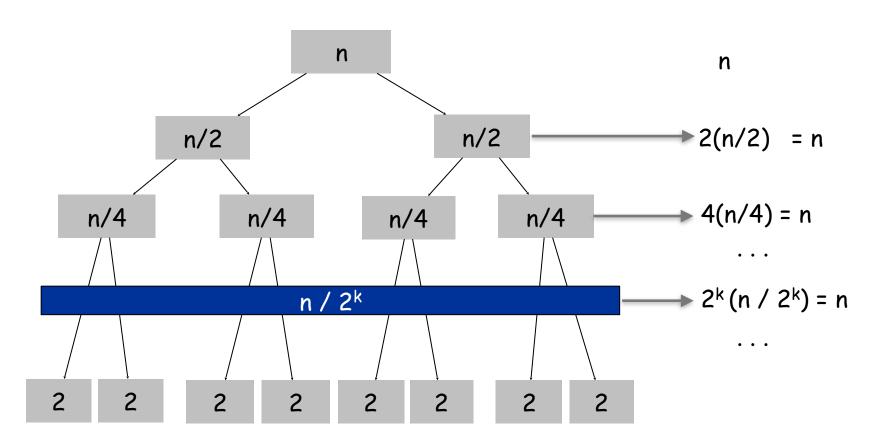
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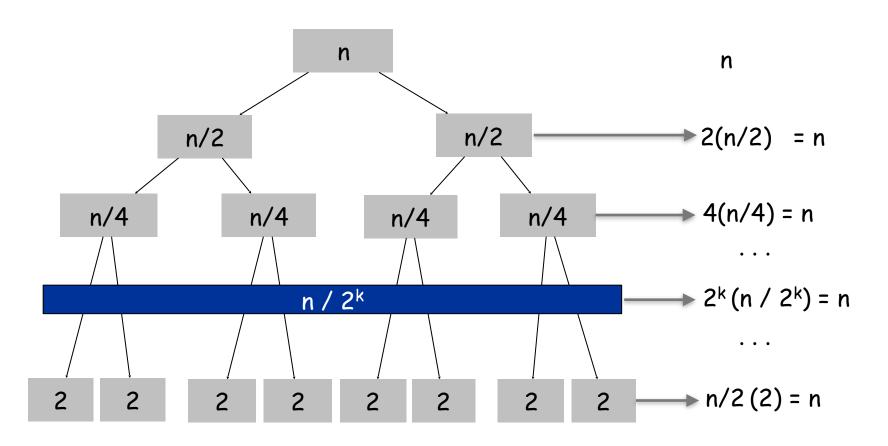
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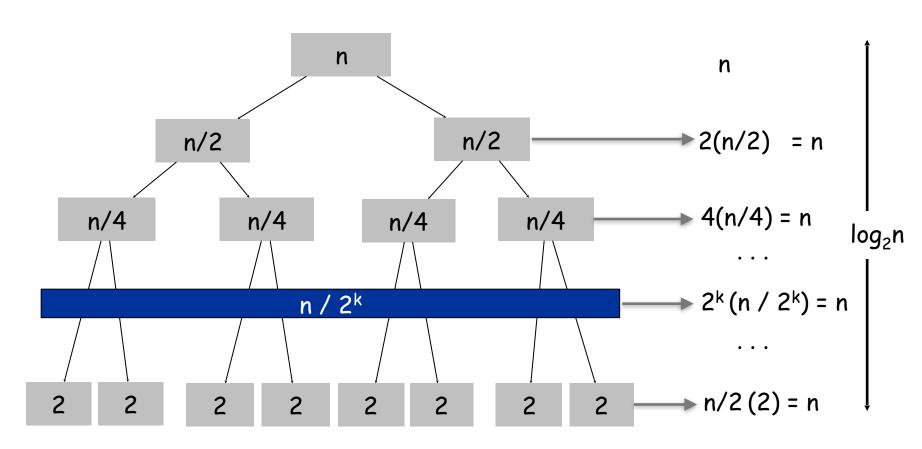
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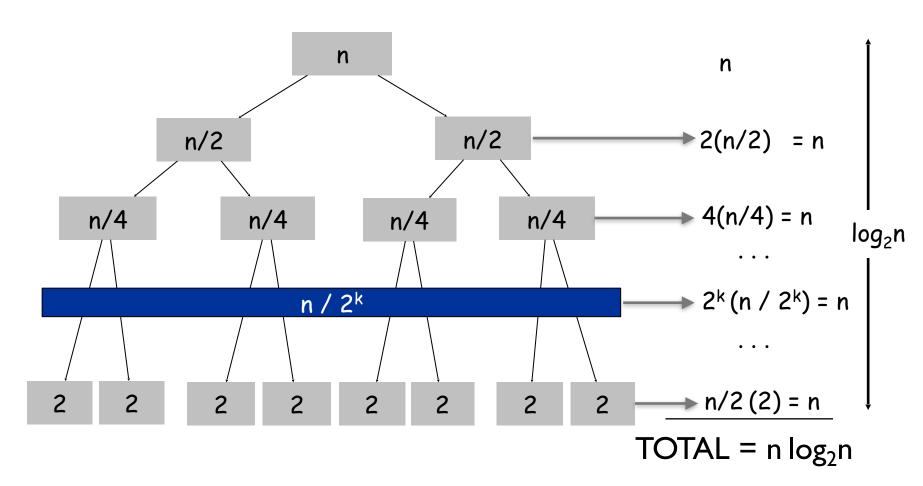
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Plan for Today

Divide and Conquer Merge-sort

Master theorem

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Divide and Conquer

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Master theorem: Generic method for solving recurrences.

Goal: Solve common divide-and-conquer recurrences:

$$T(n) = a T\left(\frac{n}{b}\right) + f(n)$$

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Terms.

- $a \ge 1$ is the number of subproblems.
- b > 0 is the factor by which subproblem size decreases.
- f(n) = work to divide/merge subproblems.