

CS170: Lecture 2

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Last Time: Place value is democratizing!

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Like the printing press!

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Like the printing press!
Reading, writing, arithmetic!

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Like the printing press!

Reading, writing, arithmetic!

Input size/representation really matters!

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Today: Chapter 2.

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Today: Chapter 2.

Divide and Conquer

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Today: Chapter 2.

Divide and Conquer \equiv Recursive.

CS170: Lecture 2

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Today: Chapter 2.

Divide and Conquer \equiv Recursive.

Lecture in one minute!

Integer Multiplication: Gauss plus recursion is magic!

$$O(n^2) \rightarrow O(n^{\log_2 3}) \approx O(n^{1.58..})$$

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Integer Multiplication: Gauss plus recursion is magic!

$$O(n^2) \rightarrow O(n^{\log_2 3}) \approx O(n^{1.58..})$$

Double size, time grows by a factor of 3.

Master's theorem: understand the recursion tree!

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

Branching by a

diminishing by b

working by $O(f(n))$.

Leaves: $n^{\log_b a}$, Work: $\sum_i a^i f\left(\frac{n}{b^i}\right)$.

Lecture in one minute!

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Recursive (Divide and Conquer) Matrix Multiplication:

8 subroutine calls of size $n/2 \times n/2$

$$\rightarrow O(n^3).$$

Strassen:

7 subroutine calls of size $n/2 \times n/2$

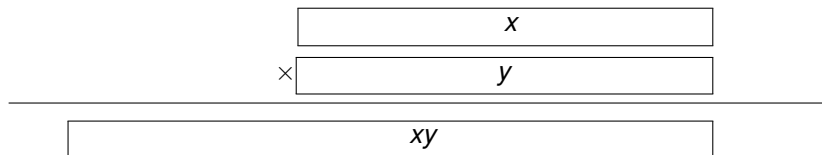
$$\rightarrow O(n^{\log_2 7}) \approx O(n^{2.8}).$$

Chapter 2.

Divide and conquer.

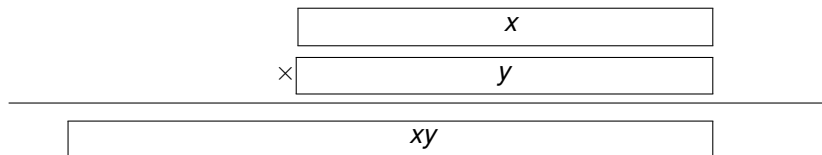
Definition of Multiplication.

n -bit numbers: x , y .



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k th "place" of xy :

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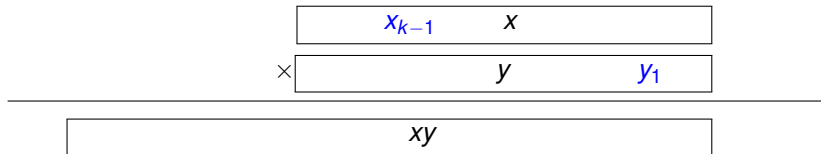
$$\begin{array}{r} \\ \\ \times \\ \hline \end{array}$$

The diagram illustrates the multiplication of two numbers, x and y . The number x is represented by a box containing x_k and x . The number y is represented by a box containing y and y_0 . A multiplication sign (\times) is placed to the left of the y box. A horizontal line is drawn below the y box. Below the line, a box contains the result xy .

k th “place” of xy : coefficient of 2^k :

Definition of Multiplication.

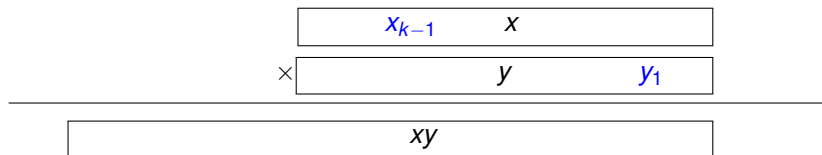
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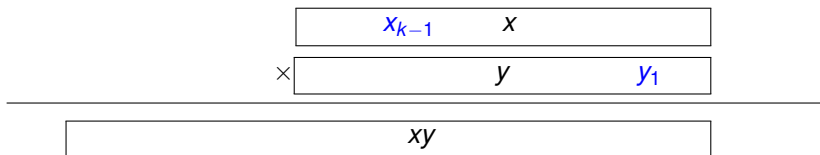


k th “place” of xy : coefficient of 2^k :

$$a_k = \sum_{i \leq k} x_i y_{k-i}.$$

Definition of Multiplication.

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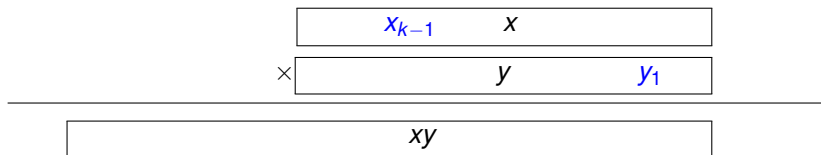
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$$a_k = \sum_{i \leq k} x_i y_{k-i}.$$

$$x * y = \sum_{k=0}^{2n} 2^k a_k.$$

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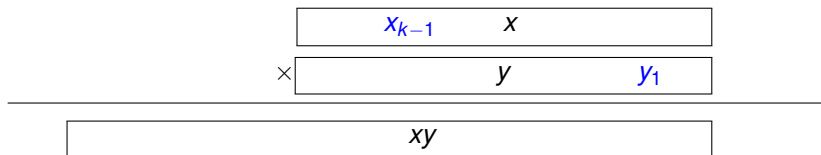
$$a_k = \sum_{i \leq k} x_i y_{k-i}.$$

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Number of “basic operations”:

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Number of “basic operations”:

$$\sum_{k \leq 2n} \min(k, 2n - k) = \Theta(n^2).$$

Recursive Algorithm for Multiplication.

Two n -bit numbers: x , y .

Recursive Algorithm for Multiplication.

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$$x = \boxed{\begin{array}{|c|c|} \hline X_L & X_R \\ \hline \end{array}}$$

Recursive Algorithm for Multiplication.

Two n -bit numbers: x , y .

$$x = \boxed{\begin{array}{|c|c|} \hline x_L & x_R \\ \hline \end{array}} = 2^{n/2}x_L + x_R$$

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Two n -bit numbers: x , y .

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$$y = \begin{array}{|c|c|} \hline y_L & y_R \\ \hline \end{array}$$

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Multiplying out

Recursive Algorithm for Multiplication.

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Multiplying out

$$x \times y$$

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Multiplying out

$$x \times y = (2^{n/2}x_L + x_R)(2^{n/2}y_L + y_R)$$

Recursive Algorithm for Multiplication.

Two n -bit numbers: x , y .

$$\begin{aligned}x &= \boxed{\begin{array}{|c|c|} \hline x_L & x_R \\ \hline \end{array}} = 2^{n/2}x_L + x_R \\y &= \boxed{\begin{array}{|c|c|} \hline y_L & y_R \\ \hline \end{array}} = 2^{n/2}y_L + y_R\end{aligned}$$

Multiplying out

$$\begin{aligned}x \times y &= (2^{n/2}x_L + x_R)(2^{n/2}y_L + y_R) \\ &= 2^n x_L y_L + 2^{n/2}(x_L y_R + x_R y_L) + x_R y_R\end{aligned}$$

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Two n -bit numbers: x , y .

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Four $n/2$ -bit multiplications: $x_L y_L$, $x_L y_R$, $x_R y_L$, $x_R y_R$.

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Recurrence:

$$T(n) = 4T\left(\frac{n}{2}\right) + O(n)$$

Recurrence for recursive algorithm.

Recurrence:

$$T(n) = 4T\left(\frac{n}{2}\right) + \Theta(n)$$

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$T(n)$ is

- (A) $\Theta(n)$.
- (B) $\Theta(n^2)$.
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A degree 4 tree of depth $\log_2 n$.

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$$4^{\log_2 n}$$

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One for each pair of digits!

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TBH, unfolded recurrence in head. Don't remember formulas.

Demo

As number of bits double:

Demo

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Demo

As number of bits double:

Elementary School Multiply:

Demo

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$$O(n^2)$$

$$n \rightarrow 2n$$

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As number of bits double:

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$$n \rightarrow 2n$$

$$\text{Runtime: } T = cn^2 \rightarrow T' = c(2n)^2 = 4(cn^2) = 4T$$

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$$\text{Runtime: } T \rightarrow 3T.$$

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$$\text{Asymptotics: } T = cn^w \rightarrow c((2n)^w) = T' = 3T = 3(cn^w).$$

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$$\dots \rightarrow 2^w = 3.$$

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$$\dots \rightarrow 2^w = 3. \text{ or } w = \log_2 3 \approx 1.58.$$

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$$\dots \rightarrow 2^w = 3. \text{ or } w = \log_2 3 \approx 1.58.$$

$$\text{Python multiply: } O(n^{\log_2 3})$$

Much better than grade school.

Multiply Complex Numbers

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What about $(32765 + 219898i)(413764 + 511110i)$?

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Multiplication!

Faster Algorithm for Multiplication.

Two n -bit numbers: x , y .

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(A) Yes.

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Compute

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$$T(n) = 3T\left(\frac{n}{2}\right) + \Theta(n)$$

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Technically: $\frac{n}{2} + 1$ bit multiplication.

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Analysis of runtime.

Recurrence for “fast algorithm”.

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$$3^{\log_2 n}$$

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But: all digits have to multiply each other!

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- (A) $\Theta(n)$
- (B) $\Theta(n^2)$
- (C) $\Theta(n^{\log_2 3})$

(C) Idea: number of base cases is $n^{\log_2 3}$.

$$3^{\log_2 n} = (2^{\log_2 3})^{\log_2 n} = n^{\log_2 3}$$

So multiplication algorithm with ..

$$T(n) = 3T\left(\frac{n}{2}\right) + \Theta(n) = \Theta(n^{\log_2 3}) = \Theta(n^{1.58\dots})!!!!$$

But: all digits have to multiply each other!

They do! $(a + b)(c + d) = ac + ad + bc + bd$

4 products from one multiplication!

Logarithms reminder.

Exponents Quiz: $(a^b)^c = (a^c)^b$?

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Yes. $(a^b)^c$

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Yes. $(a^b)^c = a^{bc}$

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Definition of log:

Logarithms reminder.

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$$a^{\log_b n}$$

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Solving recurrences.

$$T(n) = 4T\left(\frac{n}{2}\right) + cn; \quad T(1) = c$$

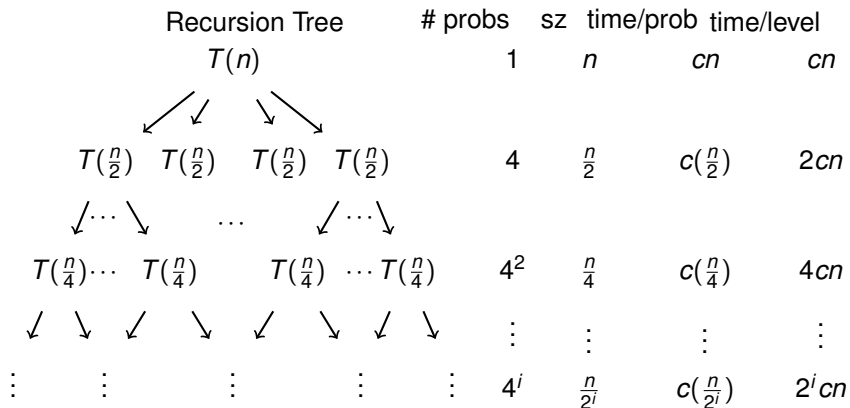
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		4^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$4cn$
		\vdots	\vdots	\vdots	\vdots
		4^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$2^i cn$

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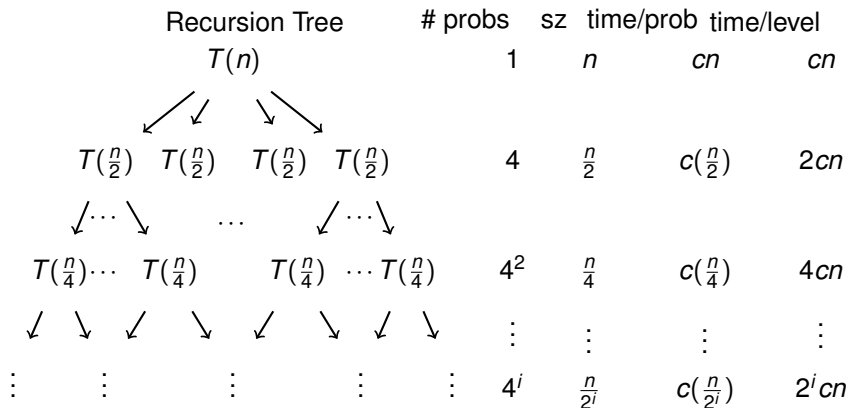
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$$\frac{n}{2^i} = 1 \text{ when } i = \log_2 n$$

Solving recurrences.

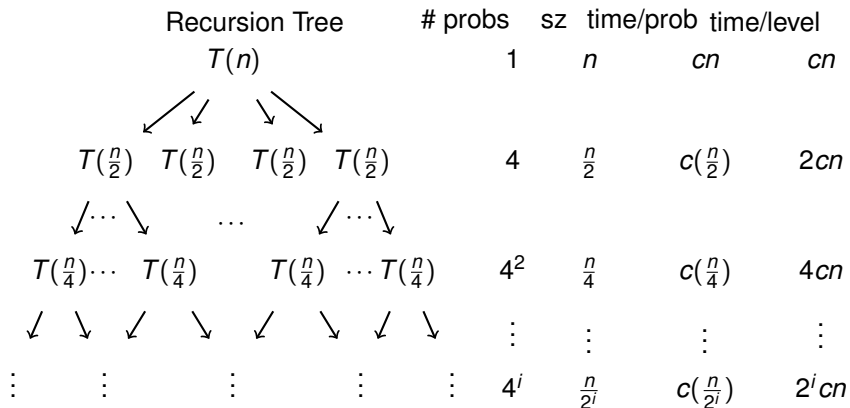
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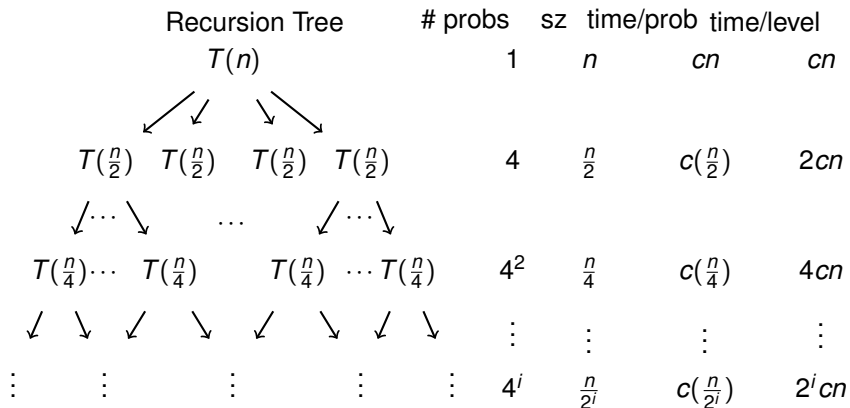


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$$4^{\log_2 n}$$

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$$4^{\log_2 n} = 2^{2 \log_2 n}$$

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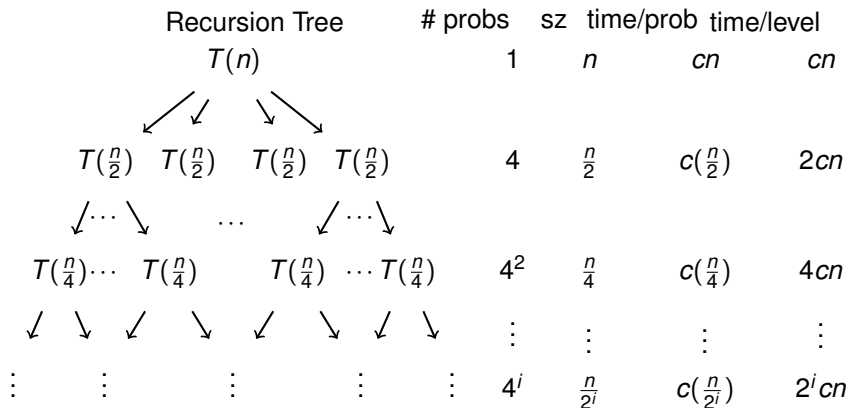
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$4^{\log n} = 2^{2 \log n} = n^2$ base case problems.

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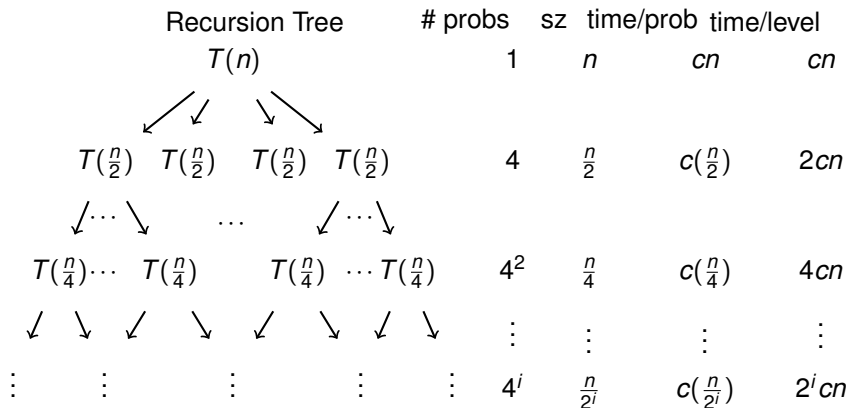
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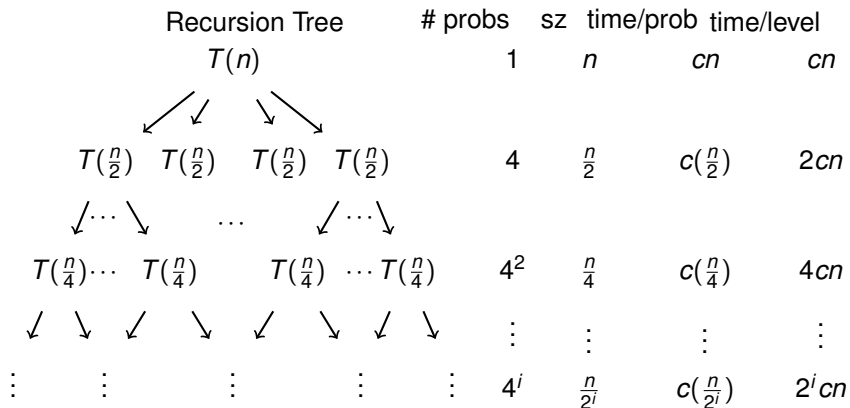
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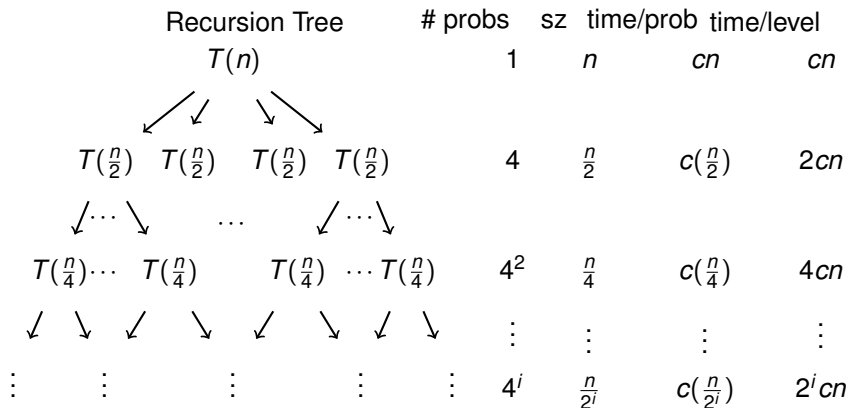
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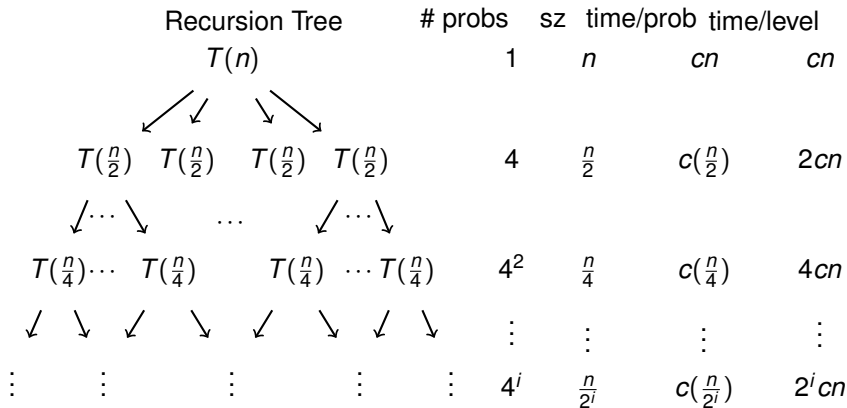
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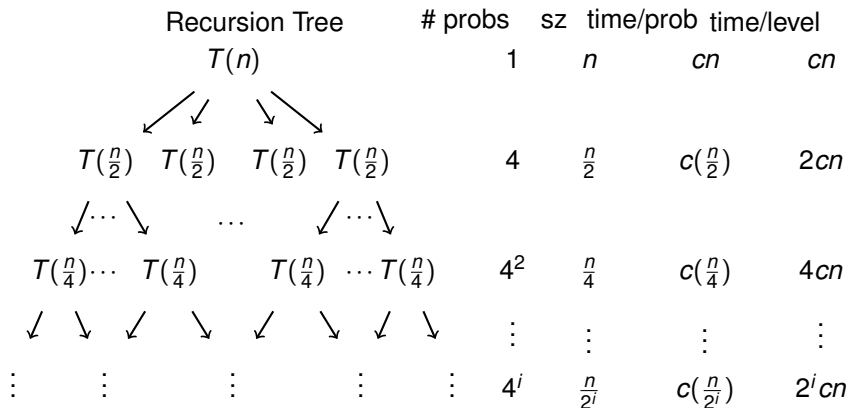
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Total Work: $cn + 2cn$

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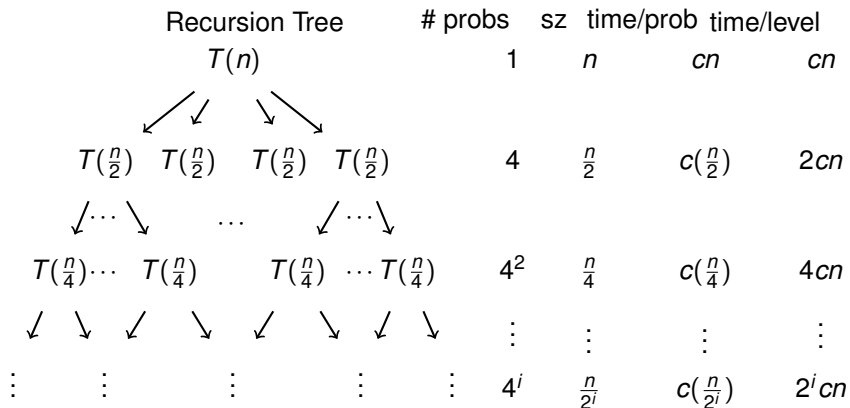
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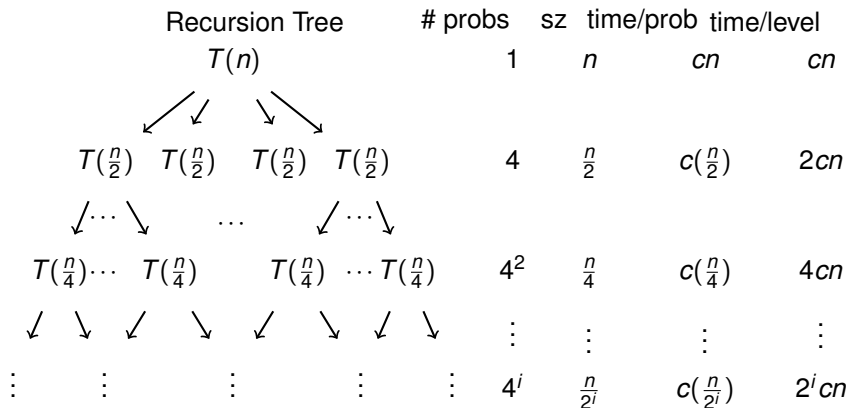
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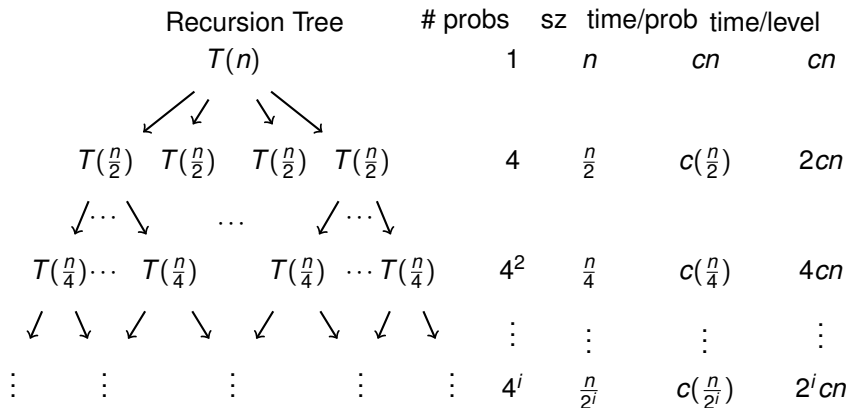
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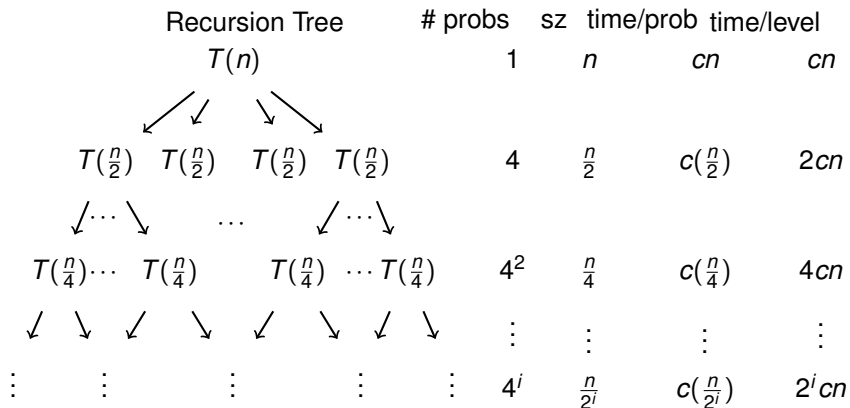
Total Work: $cn + 2cn + 4cn + \dots + cn^2 = O(n^2)$. Geometric series.

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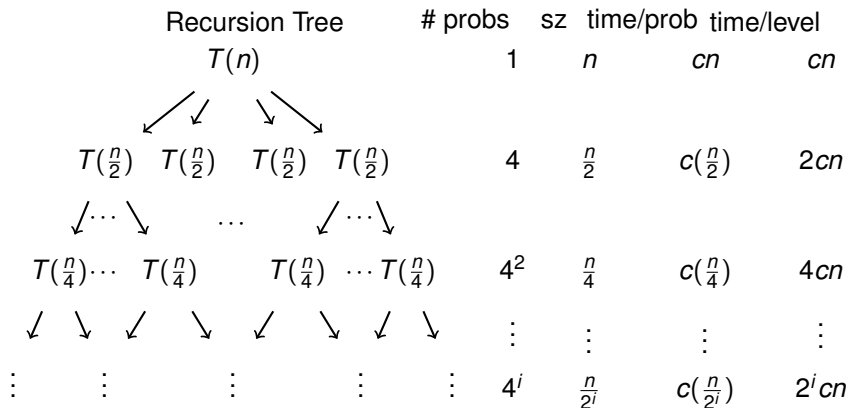
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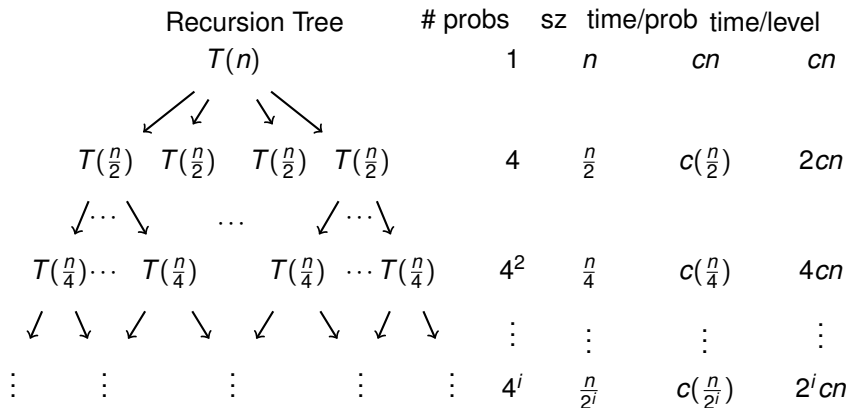
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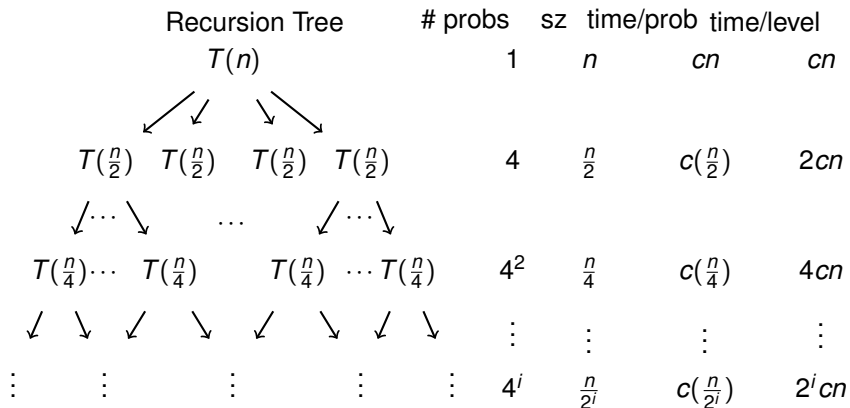
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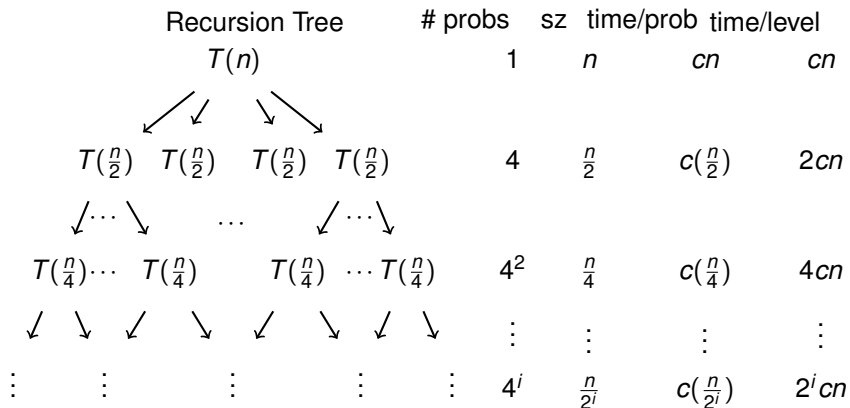


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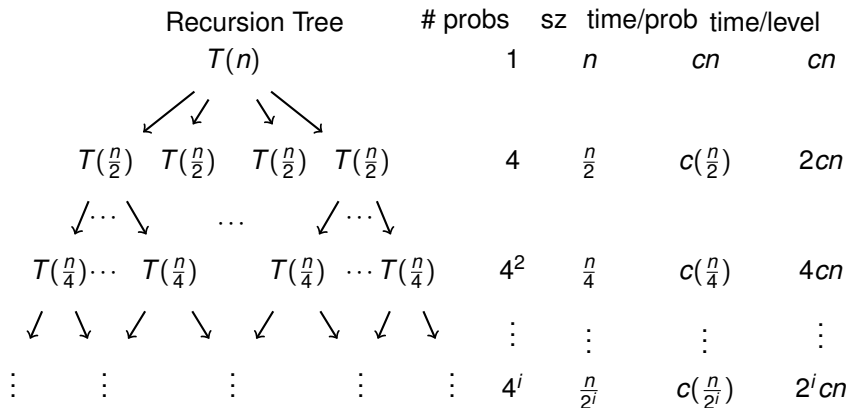
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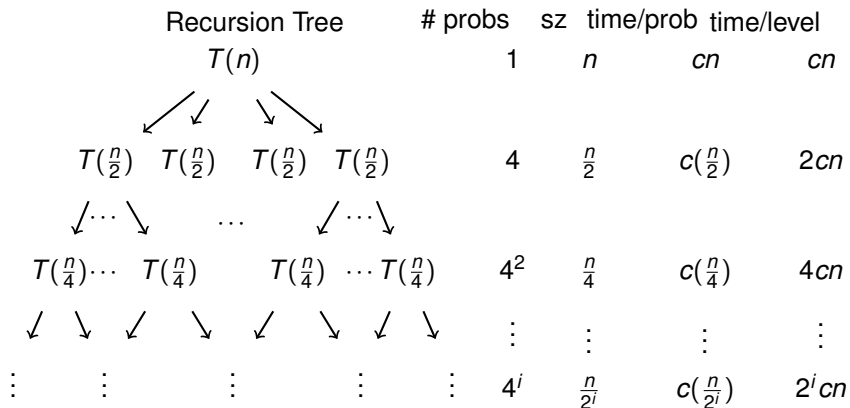
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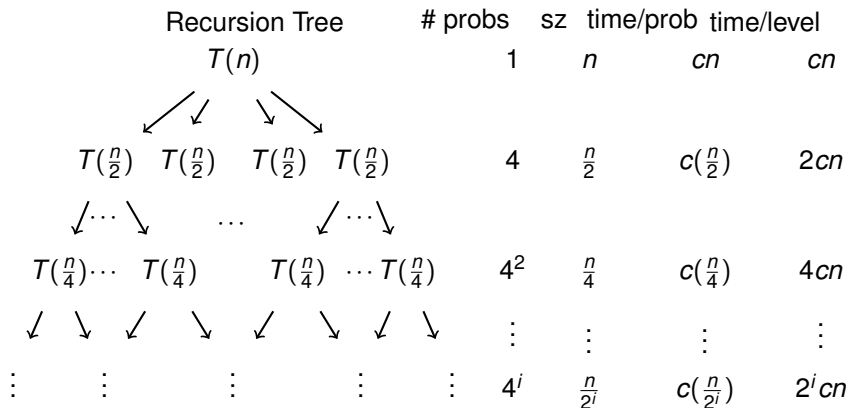
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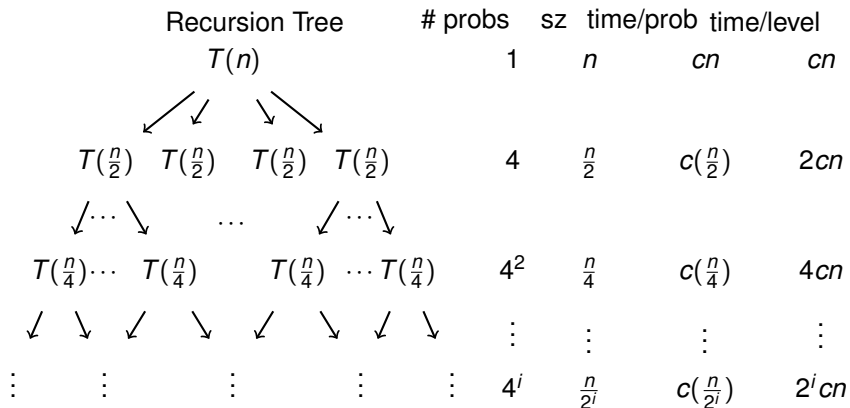
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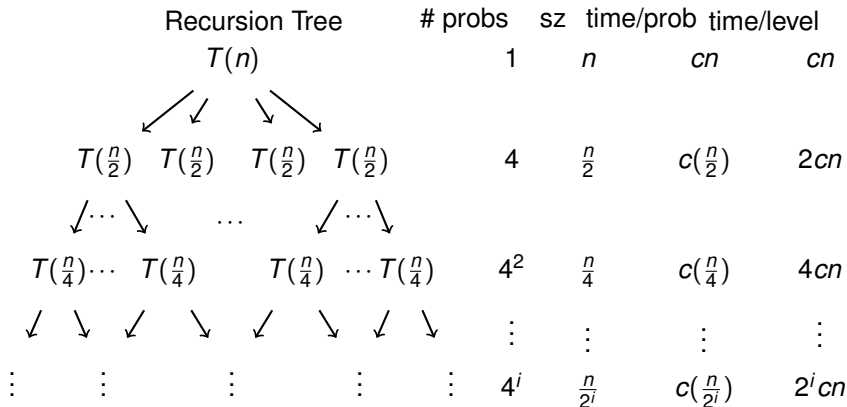
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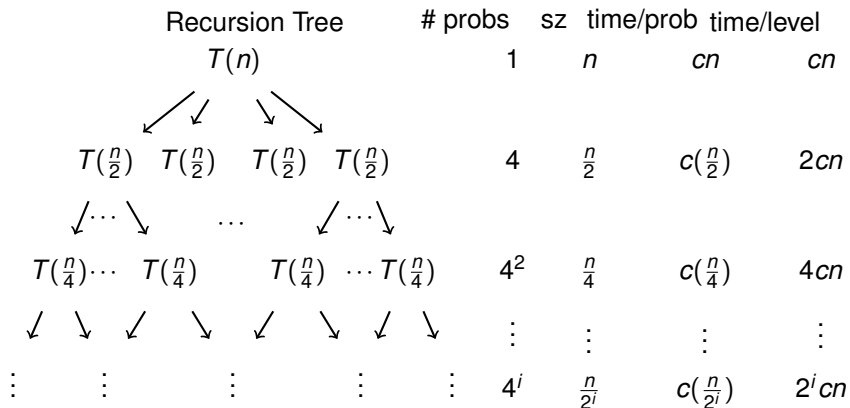
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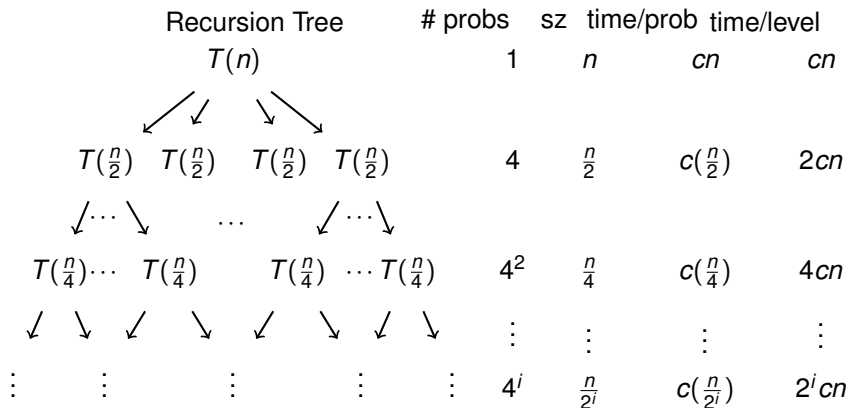
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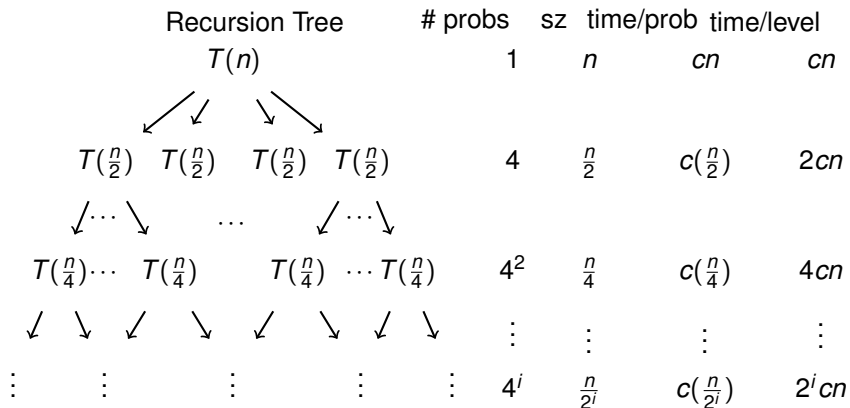
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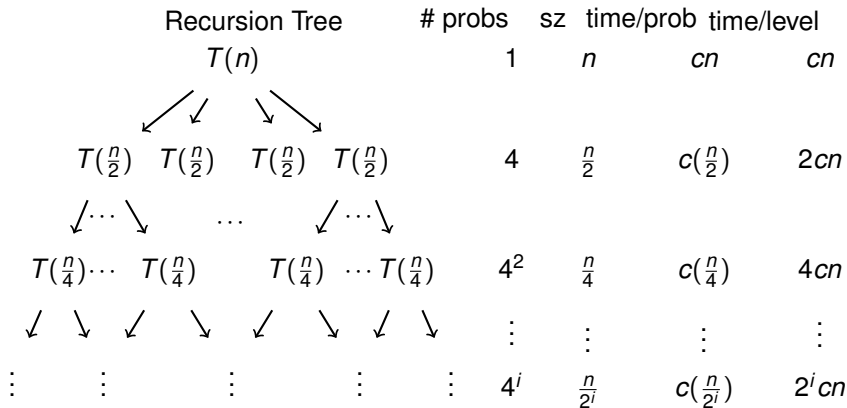
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$4^{\log_2 n} = 2^{2 \log_2 n} = n^2$ base case problems. size 1. Work/Prob: c

Work: cn^2 .

Total Work: $cn + 2cn + 4cn + \dots + cn^2 = O(n^2)$. Geometric series.

Fast multiplication.

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Recursion Tree	# probs	sz	time/prob	time/level
$T(n)$	1	n	cn	cn
$\swarrow \quad \downarrow \quad \searrow$				
$T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right)$	3	$\frac{n}{2}$	$c\left(\frac{n}{2}\right)$	$\left(\frac{3}{2}\right)cn$
$\swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow$				
$T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right) \quad T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right)$	3^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$\left(\frac{3}{2}\right)^2 cn$
$\swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow$	\vdots	\vdots	\vdots	\vdots
$\vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots$	3^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$\left(\frac{3}{2}\right)^i cn$

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Recursion Tree	# probs	sz	time/prob	time/level
$T(n)$	1	n	cn	cn
$\begin{array}{c} \swarrow \quad \downarrow \quad \searrow \\ T(\frac{n}{2}) \quad T(\frac{n}{2}) \quad T(\frac{n}{2}) \end{array}$	3	$\frac{n}{2}$	$c(\frac{n}{2})$	$(\frac{3}{2})cn$
$\begin{array}{c} \swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow \\ T(\frac{n}{4}) \cdots T(\frac{n}{4}) \quad T(\frac{n}{4}) \cdots T(\frac{n}{4}) \end{array}$	3^2	$\frac{n}{4}$	$c(\frac{n}{4})$	$(\frac{3}{2})^2 cn$
$\begin{array}{c} \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \\ \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \end{array}$	\vdots	\vdots	\vdots	\vdots
$\begin{array}{c} \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \end{array}$	3^i	$\frac{n}{2^i}$	$c(\frac{n}{2^i})$	$(\frac{3}{2})^i cn$

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$T(n)$	1	n	cn	cn
$\swarrow \quad \downarrow \quad \searrow$ $T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right)$	3	$\frac{n}{2}$	$c\left(\frac{n}{2}\right)$	$\left(\frac{3}{2}\right)cn$
$\swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow$ $T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right) \quad T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right)$	3^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$\left(\frac{3}{2}\right)^2 cn$
$\swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow$ $\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	\vdots	\vdots	\vdots	\vdots
$\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	3^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$\left(\frac{3}{2}\right)^i cn$

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$\swarrow \quad \downarrow \quad \searrow$ $T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right)$	3	$\frac{n}{2}$	$c\left(\frac{n}{2}\right)$	$\left(\frac{3}{2}\right)cn$
$\swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow$ $T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right) \quad T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right)$	3^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$\left(\frac{3}{2}\right)^2 cn$
$\swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow$ $\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	\vdots	\vdots	\vdots	\vdots
$\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	3^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$\left(\frac{3}{2}\right)^i cn$

$\frac{n}{2^i} = 1$ when $i = \log_2 n \implies$ Depth: $d = \log_2 n$.

$3^{\log_2 n}$

Fast multiplication.

$$T(n) = 3T\left(\frac{n}{2}\right) + cn; \quad T(1) = c$$

Recursion Tree	# probs	sz	time/prob	time/level
$T(n)$	1	n	cn	cn
$\swarrow \quad \downarrow \quad \searrow$ $T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right)$	3	$\frac{n}{2}$	$c\left(\frac{n}{2}\right)$	$\left(\frac{3}{2}\right)cn$
$\swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow$ $T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right) \quad T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right)$	3^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$\left(\frac{3}{2}\right)^2 cn$
$\swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow$ $\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	\vdots	\vdots	\vdots	\vdots
$\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	3^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$\left(\frac{3}{2}\right)^i cn$

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$T(n)$	1	n	cn	cn
$\begin{array}{c} \swarrow \quad \downarrow \quad \searrow \\ T(\frac{n}{2}) \quad T(\frac{n}{2}) \quad T(\frac{n}{2}) \end{array}$	3	$\frac{n}{2}$	$c(\frac{n}{2})$	$(\frac{3}{2})cn$
$\begin{array}{c} \swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow \\ T(\frac{n}{4}) \cdots T(\frac{n}{4}) \quad T(\frac{n}{4}) \cdots T(\frac{n}{4}) \end{array}$	3^2	$\frac{n}{4}$	$c(\frac{n}{4})$	$(\frac{3}{2})^2 cn$
$\begin{array}{c} \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \\ \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \end{array}$	\vdots	\vdots	\vdots	\vdots
$\begin{array}{c} \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \end{array}$	3^i	$\frac{n}{2^i}$	$c(\frac{n}{2^i})$	$(\frac{3}{2})^i cn$

$\frac{n}{2^i} = 1$ when $i = \log_2 n \implies$ Depth: $d = \log_2 n$.

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Recursion Tree	# probs	sz	time/prob	time/level
$T(n)$	1	n	cn	cn
$\begin{array}{c} \swarrow \quad \downarrow \quad \searrow \\ T(\frac{n}{2}) \quad T(\frac{n}{2}) \quad T(\frac{n}{2}) \end{array}$	3	$\frac{n}{2}$	$c(\frac{n}{2})$	$(\frac{3}{2})cn$
$\begin{array}{c} \swarrow \cdots \searrow \quad \cdots \quad \swarrow \cdots \searrow \\ T(\frac{n}{4}) \cdots T(\frac{n}{4}) \quad T(\frac{n}{4}) \cdots T(\frac{n}{4}) \end{array}$	3^2	$\frac{n}{4}$	$c(\frac{n}{4})$	$(\frac{3}{2})^2 cn$
$\begin{array}{c} \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \\ \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \end{array}$	\vdots	\vdots	\vdots	\vdots
$\begin{array}{c} \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \end{array}$	3^i	$\frac{n}{2^i}$	$c(\frac{n}{2^i})$	$(\frac{3}{2})^i cn$

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Recursion Tree	# probs	sz	time/prob	time/level
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$\swarrow \quad \downarrow \quad \searrow$ $T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right)$	3	$\frac{n}{2}$	$c\left(\frac{n}{2}\right)$	$\left(\frac{3}{2}\right)cn$
$\swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow$ $T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right) \quad T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right)$	3^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$\left(\frac{3}{2}\right)^2 cn$
$\swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow$ $\vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots$	\vdots	\vdots	\vdots	\vdots
$\vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots$	3^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$\left(\frac{3}{2}\right)^i cn$

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Recursion Tree	# probs	sz	time/prob	time/level
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$\swarrow \quad \downarrow \quad \searrow$ $T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right)$	3	$\frac{n}{2}$	$c\left(\frac{n}{2}\right)$	$\left(\frac{3}{2}\right)cn$
$\swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow$ $T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right) \quad T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right)$	3^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$\left(\frac{3}{2}\right)^2 cn$
$\swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow$ $\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	\vdots	\vdots	\vdots	\vdots
$\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	3^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$\left(\frac{3}{2}\right)^i cn$

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$\swarrow \quad \downarrow \quad \searrow$ $T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right)$	3	$\frac{n}{2}$	$c\left(\frac{n}{2}\right)$	$\left(\frac{3}{2}\right)cn$
$\swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow$ $T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right) \quad T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right)$	3^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$\left(\frac{3}{2}\right)^2 cn$
$\swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow \quad \swarrow \quad \searrow$ $\vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots$	\vdots	\vdots	\vdots	\vdots
$\vdots \quad \vdots \quad \vdots \quad \vdots \quad \vdots$	3^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$\left(\frac{3}{2}\right)^i cn$

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$\swarrow \quad \downarrow \quad \searrow$ $T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right) \quad T\left(\frac{n}{2}\right)$	3	$\frac{n}{2}$	$c\left(\frac{n}{2}\right)$	$\left(\frac{3}{2}\right)cn$
$\swarrow \cdots \searrow \quad \dots \quad \swarrow \cdots \searrow$ $T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right) \quad T\left(\frac{n}{4}\right) \cdots T\left(\frac{n}{4}\right)$	3^2	$\frac{n}{4}$	$c\left(\frac{n}{4}\right)$	$\left(\frac{3}{2}\right)^2 cn$
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$\vdots \quad \quad \quad \vdots \quad \quad \quad \vdots \quad \quad \quad \vdots$	3^i	$\frac{n}{2^i}$	$c\left(\frac{n}{2^i}\right)$	$\left(\frac{3}{2}\right)^i cn$

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$3^{\log_2 n} = n^{\log_2 3}$ base case problems. size 1. Work/Prob: c . Work: $cn^{\log_2 3}$.

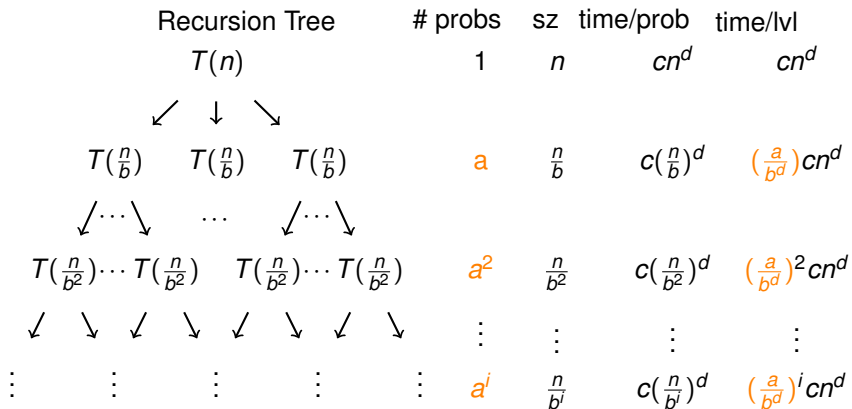
Total Work: $cn + \left(\frac{3}{2}\right)cn + \dots + cn^{\log_2 3} = O(n^{\log_2 3})$ Geometric series.

Divide and Conquer: In general.

$$T(n) = aT\left(\frac{n}{b}\right) + O(n^d); \quad T(1) = c$$

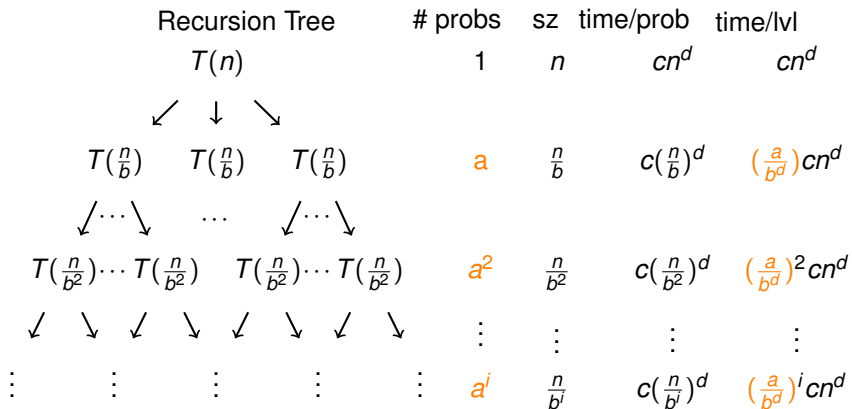
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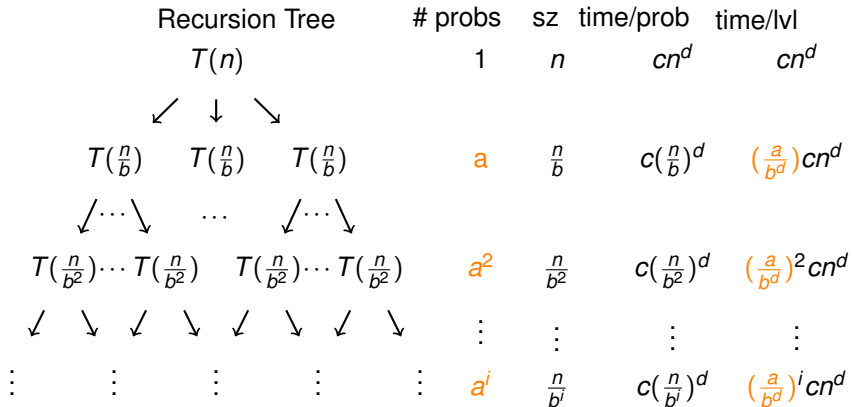
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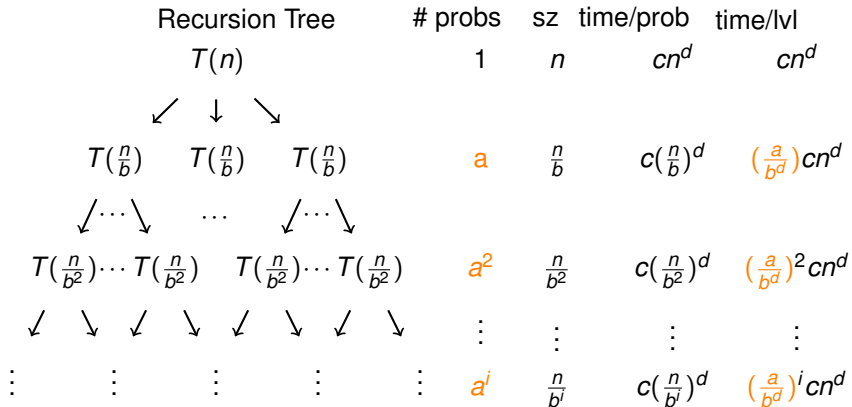
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Level i work: $\left(\frac{a}{b^d}\right)^i n^d$.

Master's Theorem

Depth: $\log_b n$.

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Depth: $\log_b n$.

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Geometric series:

Master's Theorem

Depth: $\log_b n$.

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Geometric series: If $\frac{a}{b^d} < 1$ ($d > \log_b a$), first term dominates

$$O(n^d),$$

Master's Theorem

Depth: $\log_b n$.

Level i work:

$$\left(\frac{a}{b^d}\right)^i n^d.$$

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Geometric series: If $\frac{a}{b^d} < 1$ ($d > \log_b a$), first term dominates

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if $\frac{a}{b^d} > 1$ ($d < \log_b a$), last term dominates.

$$O(n^{\log_b a}),$$

Master's Theorem

Depth: $\log_b n$.

Level i work:

$$\left(\frac{a}{b^d}\right)^i n^d.$$

Total:

$$n^d \sum_{i=0}^{\log_b n} \left(\frac{a}{b^d}\right)^i$$

Geometric series: If $\frac{a}{b^d} < 1$ ($d > \log_b a$), first term dominates

$$O(n^d),$$

if $\frac{a}{b^d} > 1$ ($d < \log_b a$), last term dominates.

$$O(n^{\log_b a}),$$

and if $\frac{a}{b^d} = 1$ ($d = \log_b a$), then all terms are the same

$$O(n^d \log_b n).$$

Master's Theorem: examples.

For a recurrence $T(n) = aT(n/b) + O(n^d)$

We have

$$d > \log_b a \quad T(n) = O(n^d)$$

$$d < \log_b a \quad T(n) = O(n^{\log_b a})$$

$$d = \log_b a \quad T(n) = O(n^d \log_b n).$$

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$$T(n) = 4T\left(\frac{n}{2}\right) + O(n) \quad a = 4, b = 2, \text{ and } d = 1.$$

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$T(n) = 4T(\frac{n}{2}) + O(n)$ $a = 4$, $b = 2$, and $d = 1$.

$$d = 1 < 2 = \log_2 4 = \log_b a$$

Master's Theorem: examples.

For a recurrence $T(n) = aT(n/b) + O(n^d)$

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$$d = \log_b a \quad T(n) = O(n^d \log_b n).$$

$T(n) = 4T(\frac{n}{2}) + O(n)$ $a = 4$, $b = 2$, and $d = 1$.

$d = 1 < 2 = \log_2 4 = \log_b a \implies T(n) = O(n^{\log_b a}) = O(n^2)$.

Master's Theorem: examples.

For a recurrence $T(n) = aT(n/b) + O(n^d)$

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Strassen

Matrix multiplication.

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Strassen: Divide! conquer!

Matrix Multiplication

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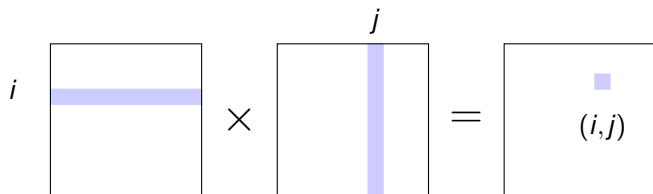
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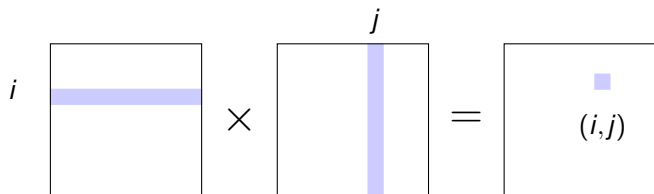


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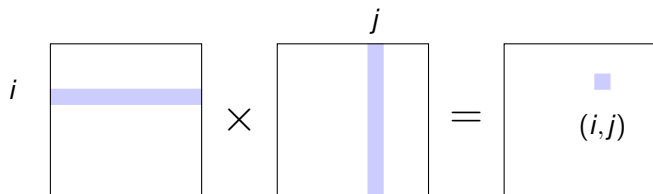
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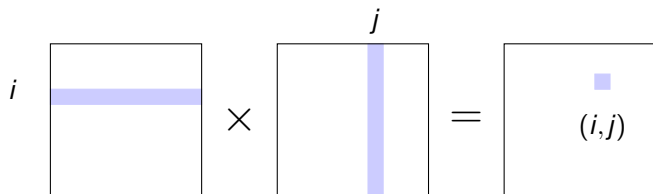
Runtime?

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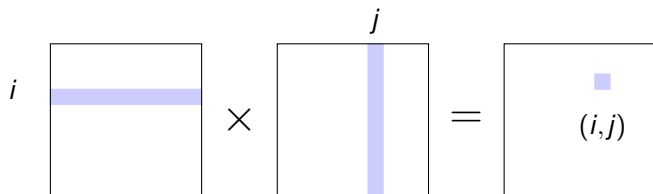
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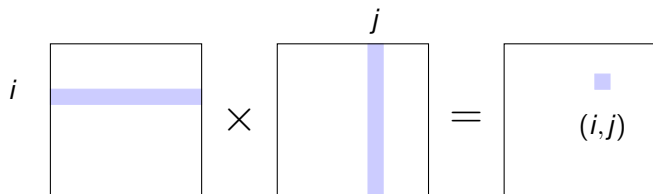
Runtime? $O(n^2)$? $O(n^3)$?

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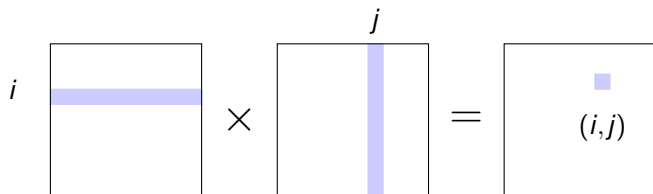
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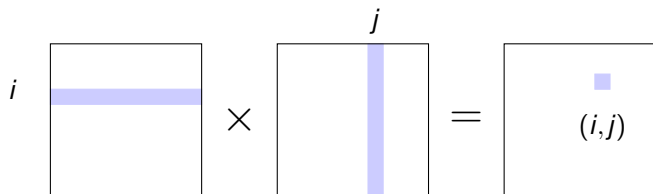
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Divide and Conquer

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} \begin{bmatrix} E & F \\ G & H \end{bmatrix} = \begin{bmatrix} & \\ & \end{bmatrix}$$

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Masters: $O(n^{\log_2 8}) = O(n^3)$.

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$$P_1 = A(F - H) \quad P_5 = (A + D)(E + H)$$

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7 multiplies! Recurrence?

$$T(n) = 7T\left(\frac{n}{2}\right) + O(n^2)$$

From Masters:

(A) $O(n^2)$? (B) $O(n^{\log_2 7} \log n)$? (C) $T(n) = O(n^{\log_2 7})$?

Leaf subproblems dominate runtime!

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Commonly used in practice!

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Strassen's 7 multiplies + recursion \implies faster matrix multiplication.

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