

COL 351 : ANALYSIS & DESIGN OF ALGORITHMS

LECTURE 6

DIVIDE & CONQUER I :

PROOF AND APPLICATIONS OF MASTER THEOREM

AUG 02, 2024

|

ROHIT VAISH

LAST TIME

Counting Inversions

Matrix multiplication

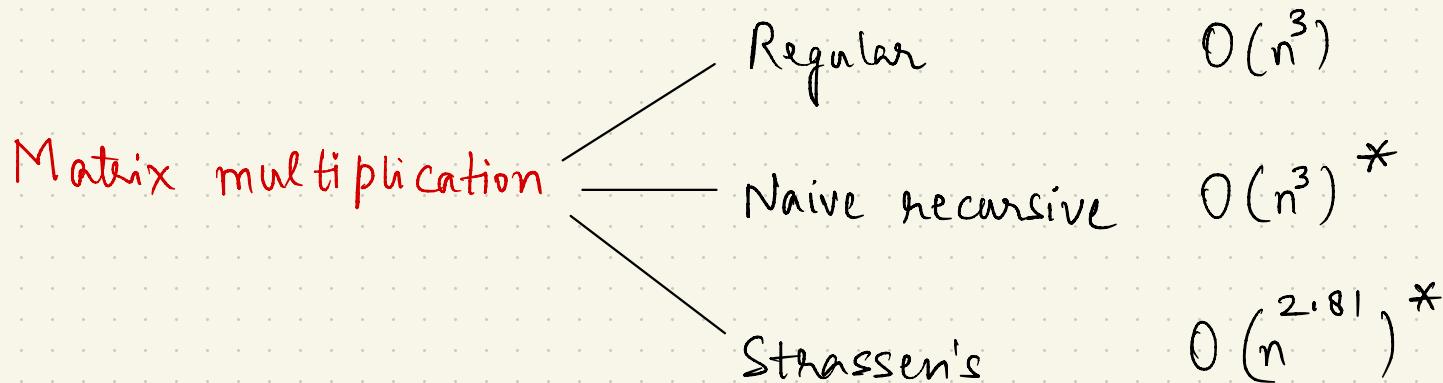
LAST TIME

Counting Inversions via enhanced Merge Sort $O(n \log n)$

Matrix multiplication

LAST TIME

Counting Inversions via enhanced Merge Sort $O(n \log n)$



MASTER THEOREM

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Theorem: If $T(n) \leq aT\left(\frac{n}{b}\right) + O(n^d)$, then

$$T(n) = \left\{ \begin{array}{l} \dots \\ \dots \\ \dots \end{array} \right.$$

Case 1

Case 2

Case 3

MASTER THEOREM

Theorem: If $T(n) \leq aT\left(\frac{n}{b}\right) + O(n^d)$, then

$$T(n) = \begin{cases} & \text{if } a = b^d \quad \text{Case 1} \\ & \text{if } a < b^d \quad \text{Case 2} \\ & \text{if } a > b^d \quad \text{Case 3} \end{cases}$$

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Theorem: If $T(n) \leq aT\left(\frac{n}{b}\right) + O(n^d)$, then

$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \quad \text{Case 1} \\ O(n^d) & \text{if } a < b^d \quad \text{Case 2} \\ O(n^{\log_b a}) & \text{if } a > b^d \quad \text{Case 3} \end{cases}$$

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only upper bounds

EXAMPLES

EXAMPLES

Merge Sort

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Merge Sort

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

maximum number of operations

the algorithm needs for an
input of size n

EXAMPLES

Merge Sort

$$a=2, b=2, d=1$$

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

EXAMPLES

Merge Sort

$$a=2, b=2, d=1$$

(Case 1)

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Recall $T(n) = \begin{cases} O(n^d \log n) & \text{if } a=b^d \\ O(n^d) & \text{if } a < b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$

Case 1
Case 2
Case 3

EXAMPLES

Merge Sort

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$$a=2, b=2, d=1$$

(Case 1)

$O(n \log n)$

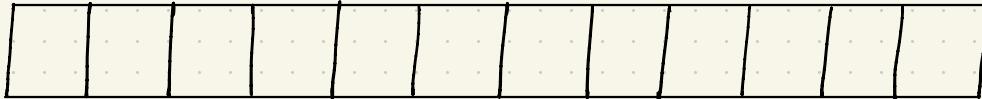
BINARY SEARCH

input: a sorted array A of length n and a target T
output: index of T in A (if exists)

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Look for "14"



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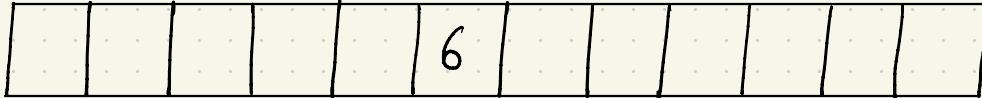
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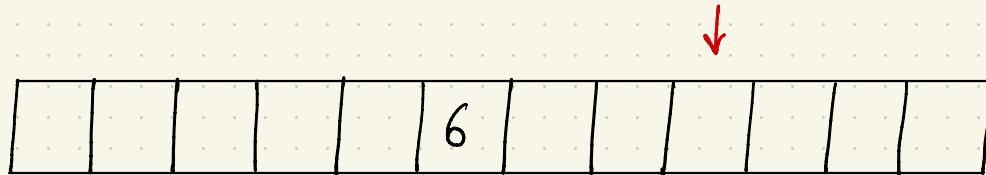
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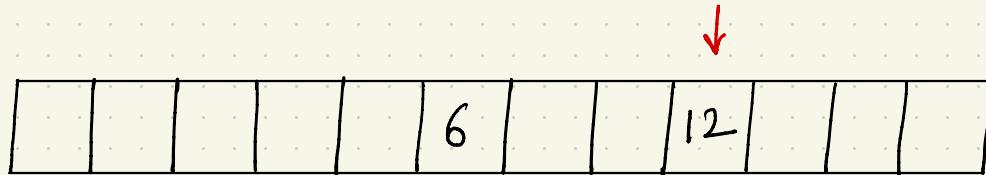
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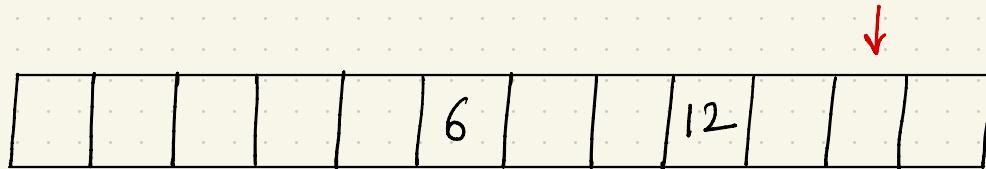
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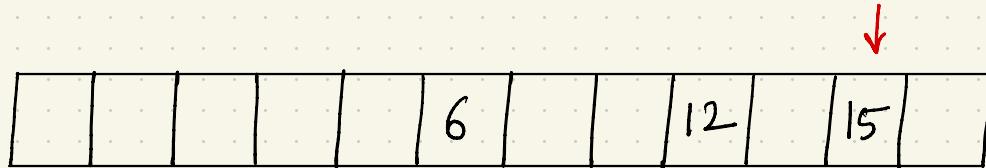
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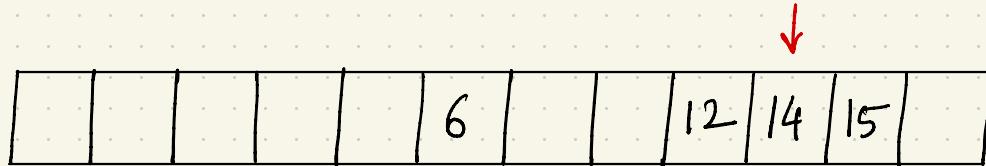
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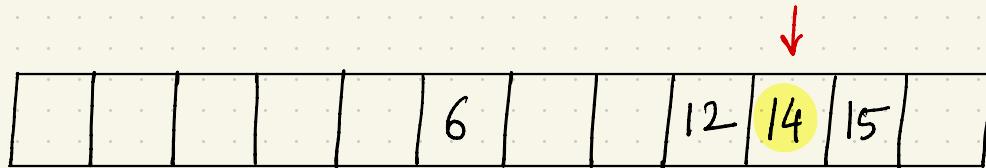
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BINARY SEARCH

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Look for "14"



BINARY SEARCH

input: a sorted array A of length n and a target T
output: index of T in A (if exists)

initialize $l := 0$

initialize $h := n - 1$

initialize $m := \left\lfloor \frac{l + h}{2} \right\rfloor$

BINARY SEARCH

input: a sorted array A of length n and a target T
output: index of T in A (if exists)

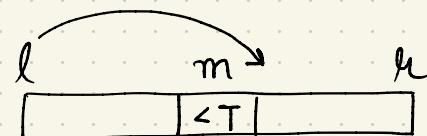
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if $A[m] < T$

because with $l := m + 1$



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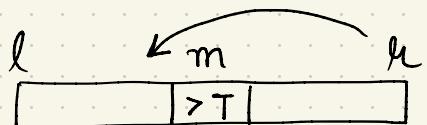
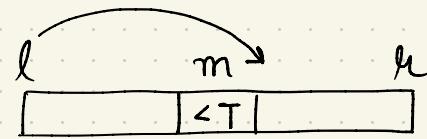
initialize $m := \left\lfloor \frac{l + h}{2} \right\rfloor$

if $A[m] < T$

 because with $l := m + 1$

else if $A[m] > T$

 because with $h := m - 1$



BINARY SEARCH

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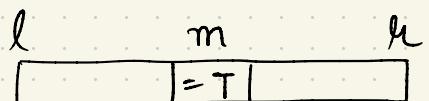
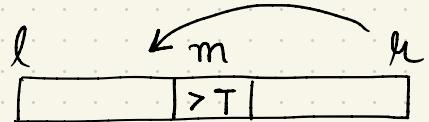
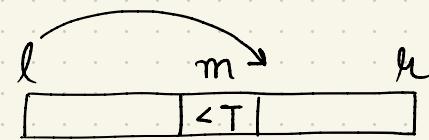
else if $A[m] > T$

 recuse with $h := m - 1$

else

 return m

return NO



EXAMPLES

Merge Sort

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

$$a=2, b=2, d=1$$

(Case 1)

$$O(n \log n)$$

Binary Search

$$a=1, b=2, d=0$$

$$T(n) \leq T\left(\frac{n}{2}\right) + O(1)$$

EXAMPLES

Merge Sort

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

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EXAMPLES

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Binary Search

$$T(n) \leq T\left(\frac{n}{2}\right) + O(1)$$

$$a=1, b=2, d=0$$

(Case 1)

$$O(\log n)$$

don't need to read
the entire input to
solve the problem

RECURSIVE ALGORITHM FOR INTEGER MULTIPLICATION

input: two n digit numbers x and y

output: the product $x \cdot y$

if $n = 1$

return $x \cdot y$

else

write $x = 10^{\frac{n}{2}} \cdot a + b$ and $y = 10^{\frac{n}{2}} \cdot c + d$

return $10^n ac + 10^{\frac{n}{2}}(ad + bc) + bd$

EXAMPLES

Recursive integer multⁿ

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

EXAMPLES

Recursive integer multⁿ

$$a = 4, b = 2, d = 1$$

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EXAMPLES

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$$a = 4, b = 2, d = 1$$

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

(Case 3)

Recall $T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \quad \text{Case 1} \\ O(n^d) & \text{if } a < b^d \quad \text{Case 2} \\ O(n^{\lg_b a}) & \text{if } a > b^d \quad \text{Case 3} \end{cases}$

EXAMPLES

Recursive integer multⁿ

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

$$a = 4, b = 2, d = 1$$

(Case 3)

$$O(n^{\log_2 4}) = O(n^2)$$

Same as
grade-school algorithm

KARATSUBA MULTIPLICATION

①

Compute $a \cdot c = 56 \times 12 = 672$

$$x = \begin{matrix} a \\ 5 & 6 \end{matrix} \quad y = \begin{matrix} b \\ 7 & 8 \\ 1 & 2 \\ 3 & 4 \\ c & d \end{matrix}$$

②

compute $b \cdot d = 78 \times 34 = 2652$

③

compute $(a+b) \cdot (c+d) = 134 \times 46 = 6164$

④

compute $③ - ② - ① = 2840$

⑤

Compute $672 \ 0000$ from ①

$$2652 \quad \text{from } ②$$

$$2840 \ 00 \quad \text{from } ④$$

$$\begin{array}{r} 672 \ 0000 \\ 2652 \\ \hline 7006652 = x \cdot y \end{array}$$

EXAMPLES

Recursive integer multⁿ

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$$a = 4, b = 2, d = 1$$

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$$O(n^{\log_2 4}) = O(n^2)$$

Same as
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Karatsuba multⁿ

$$T(n) \leq 3T\left(\frac{n}{2}\right) + O(n)$$

$$a = 3, b = 2, d = 0$$

EXAMPLES

Recursive integer multⁿ

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

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Same as
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$$a = 3, b = 2, d = 0$$

(Case 3)

$$O\left(n^{\log_2 3}\right) = O(n^{1.59})$$

faster!

EXAMPLES

Recursive integer multⁿ

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Same as
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$$a=3, b=2, d=0$$

(Case 3)

$$O\left(n^{\log_2 3}\right) = O(n^{1.59})$$

faster!

Python switches from grade-school to karatsuba at $n=70$.

RECURSIVE MATRIX MULTIPLICATION

$$X = \left[\begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]_{n \times n}$$

$$Y = \left[\begin{array}{c|c} E & F \\ \hline G & H \end{array} \right]_{n \times n}$$

Step 1 : Recursively compute the eight matrix products

$$X \cdot Y = \left[\begin{array}{cc} AE + BG & AF + BH \\ CE + DG & CF + DH \end{array} \right]$$

Step 2 : Do the necessary additions ($\Theta(n^2)$ time)

EXAMPLES

Recursive matrix multⁿ

$$a = 8, b = 2, d = 2$$

$$T(n) \leq 8T\left(\frac{n}{2}\right) + O(n^2)$$

EXAMPLES

Recursive matrix multⁿ

$$T(n) \leq 8T\left(\frac{n}{2}\right) + O(n^2)$$

$$a=8, b=2, d=2$$

(Case 3)

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

Same as the
Swim-and-dive algorithm

STRASSEN'S ALGORITHM

Step 1: Recursively compute only seven matrix products

Step 2: Do the necessary additions and subtractions ($\Theta(n^2)$ time)

$$P_1 = A \cdot (F - H)$$

$$P_2 = (A + B) \cdot H$$

$$P_3 = (C + D) \cdot E$$

$$P_4 = D \cdot (G - E)$$

$$P_5 = (A + D) \cdot (E + H)$$

$$P_6 = (B - D) \cdot (G + H)$$

$$P_7 = (A - C) \cdot (E + F)$$

$$X \cdot Y = \begin{bmatrix} P_5 + P_4 - P_2 + P_6 & P_1 + P_2 \\ P_3 + P_4 & P_1 + P_5 - P_3 - P_7 \end{bmatrix}$$

EXAMPLES

Recursive matrix multⁿ

$$T(n) \leq 8T\left(\frac{n}{2}\right) + O(n^2)$$

$$a=8, b=2, d=2$$

(Case 3)

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

Same as the
Swim-and-dive algorithm

Strassen's algorithm

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

$$a=7, b=2, d=1$$

(Case 3)

$$O(n^{\log_2 7}) = O(n^{2.81})$$

faster!

EXAMPLES

Fictitious recurrence

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

EXAMPLES

Fictitious recurrence

$$a = 2, b = 2, d = 2$$

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

Recall $T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \\ O(n^d) & \text{if } a < b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$

Case 1 Case 2 Case 3

EXAMPLES

Fictitious recurrence

$$a=2, b=2, d=2$$

$O(n^2)$

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

(Case 2)

$$\text{Recall } T(n) = \begin{cases} O(n^d \log n) & \text{if } a=b^d \\ O(n^d) & \text{if } a < b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

Case 1 Case 2 Case 3

PROOF OF MASTER THEOREM

MASTER THEOREM

Theorem: If $T(n) \leq aT\left(\frac{n}{b}\right) + O(n^d)$, then

$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \quad \text{Case 1} \\ O(n^d) & \text{if } a < b^d \quad \text{Case 2} \\ O(n^{\log_b a}) & \text{if } a > b^d \quad \text{Case 3} \end{cases}$$

Why three cases ?

Three different types of recursion trees.

Formally,

Base case: $T(1) = c$

General case: for $n > 1$

$$T(n) \leq a \cdot T\left(\frac{n}{b}\right) + c \cdot n^d$$

Formally,

Base case: $T(1) = c$

Can assume same constant

General case: for $n > 1$

$$T(n) \leq a \cdot T\left(\frac{n}{b}\right) + c \cdot n^d$$

Formally,

Base case: $T(1) = c$

General case: for $n > 1$ Can be $n > n_0$ in general

$$T(n) \leq a \cdot T\left(\frac{n}{b}\right) + c \cdot n^d$$

Formally,

Base case: $T(1) = c$

General case: for $n > 1$

$$T(n) \leq a \cdot T\left(\frac{n}{b}\right) + c \cdot n^d$$

will assume that n

is a power of b

$(n = b^k \text{ for some } k)$

Formally,

Base case: $T(1) = c$

General case: for $n > 1$

$$T(n) \leq a \cdot T\left(\frac{n}{b}\right) + c \cdot n^d$$

High level idea: Analyze using recursion trees (like merge sort)

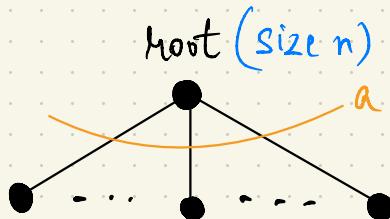
RECURSION TREE

RECURSION TREE

Root (size n)
●

level 0 outermost call

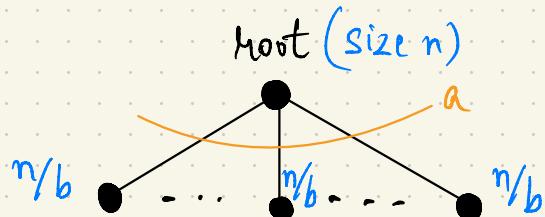
RECURSION TREE



level 0 outermost call

level 1 first batch of recursive calls

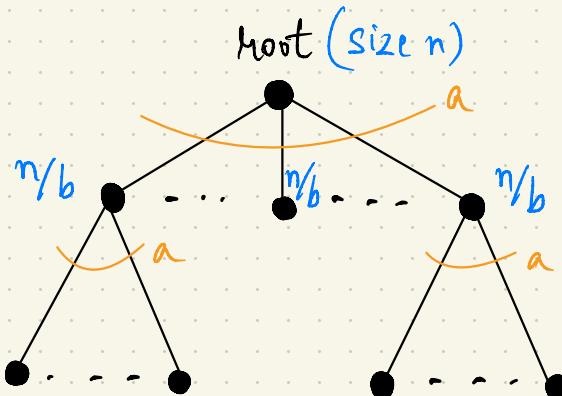
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level 0 outermost call

level 1 first batch of recursive
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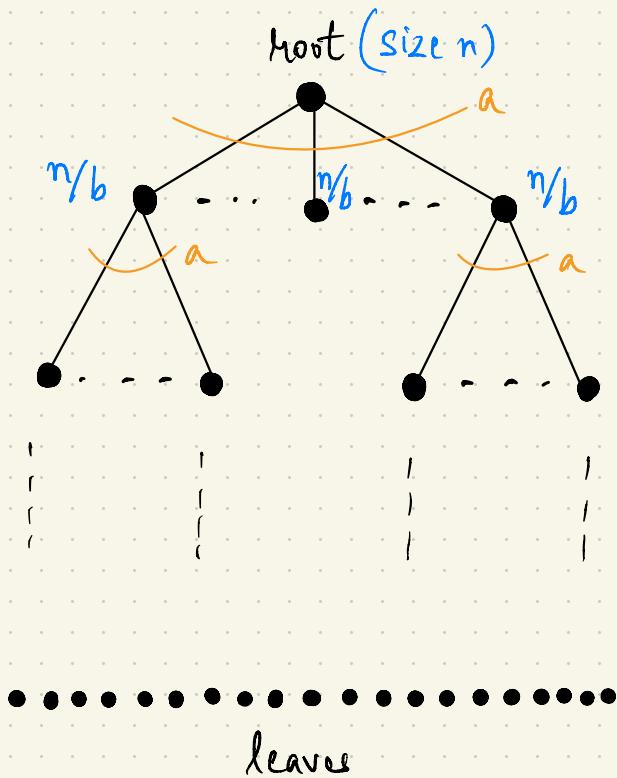
RECURSION TREE



level 0 outermost call
level 1 first batch of recursive calls

level 2 second batch

RECURSION TREE

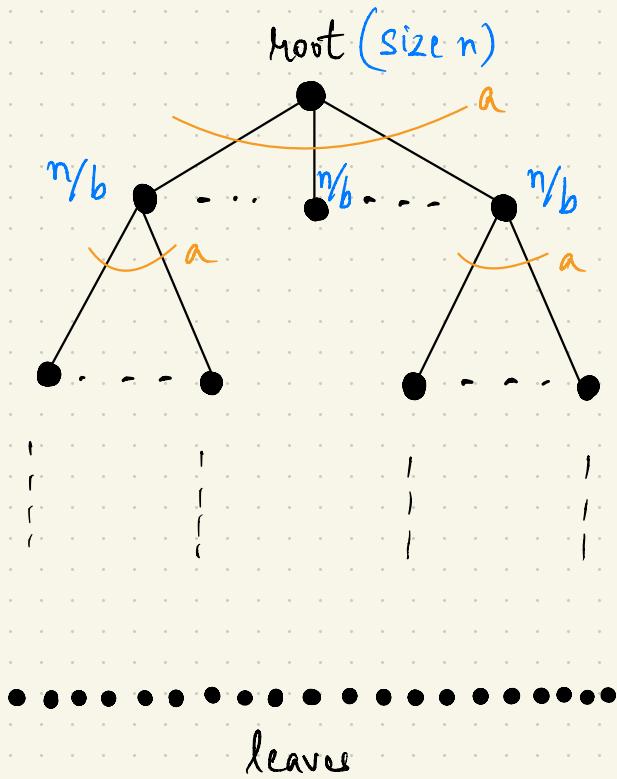


level 0 outermost call
level 1 first batch of recursive calls

level 2 second batch

level ? base cases

RECURSION TREE

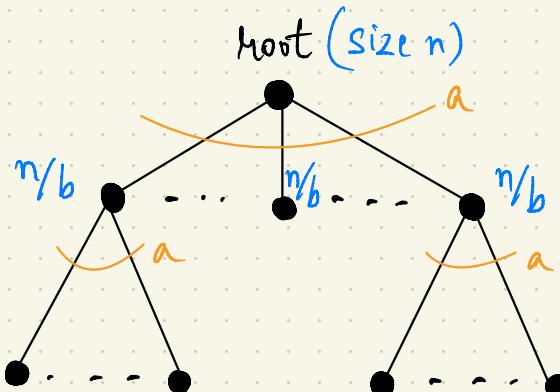


level 0 outermost call
level 1 first batch of recursive calls

level 2 second batch

level $\log_b n$ base cases
 # times n is divided by b to go below 1

RECURSION TREE



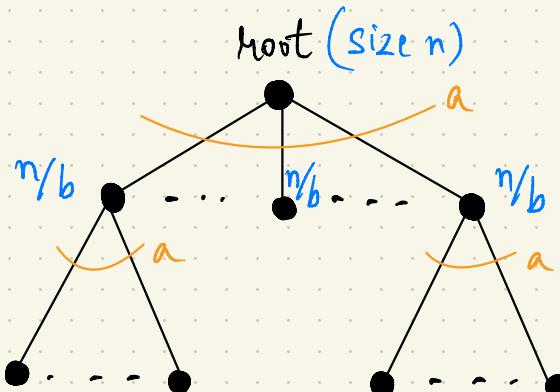
At level j

subproblems =



leaves

RECURSION TREE



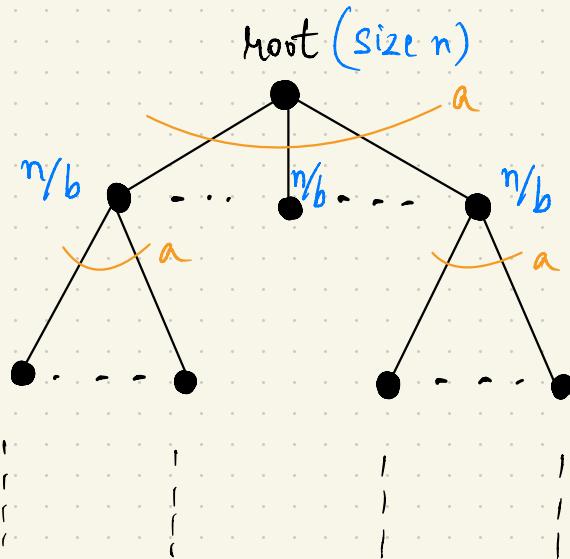
At level j

$$\# \text{ subproblems} = a^j$$



leaves

RECURSION TREE



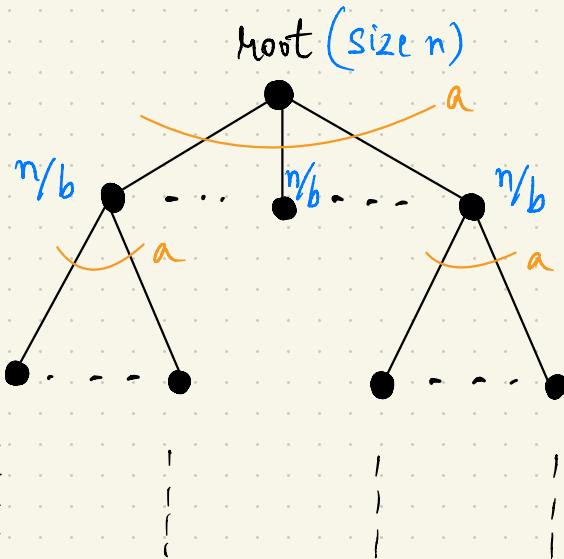
At level j

$$\# \text{ subproblems} = a^j$$

$$\text{subproblem size} =$$

.....
leaves

RECURSION TREE

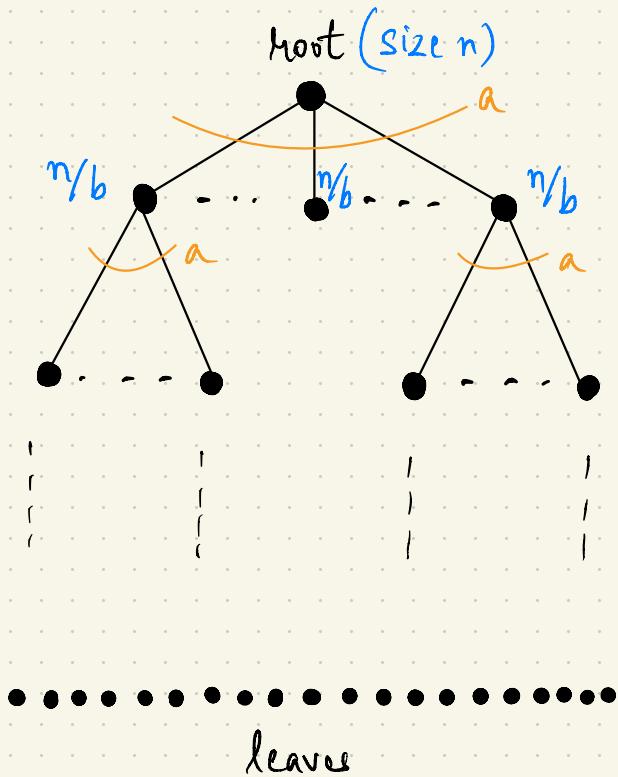


At level j

$$\# \text{ subproblems} = a^j$$

$$\text{subproblem size} = \frac{n}{b^j}$$

RECURSION TREE



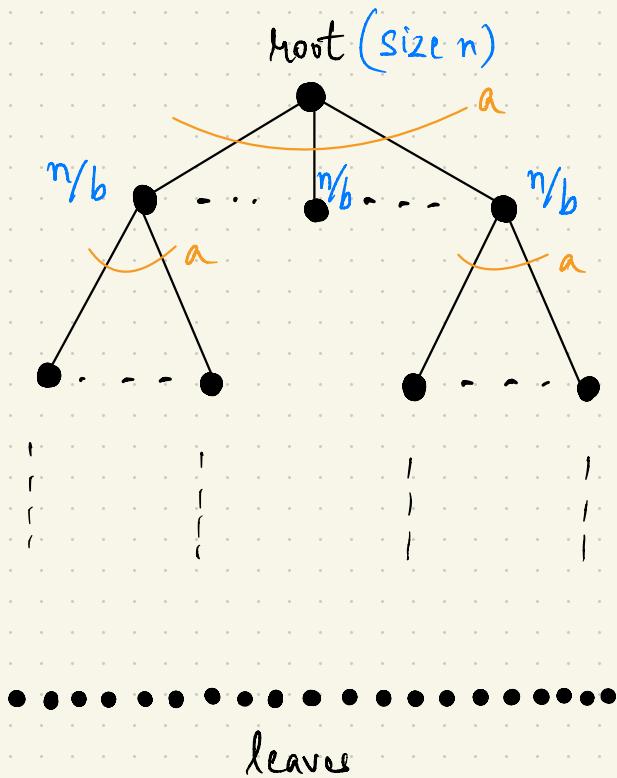
At level j

$$\# \text{ subproblems} = a^j$$

$$\text{subproblem size} = \frac{n}{b^j}$$

Total # levels =

RECURSION TREE



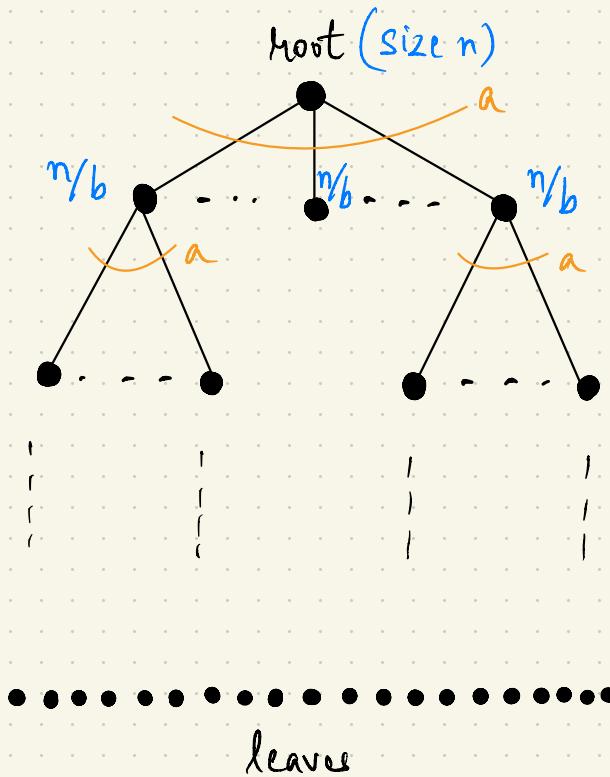
At level j

$$\# \text{ subproblems} = a^j$$

$$\text{subproblem size} = \frac{n}{b^j}$$

$$\text{Total } \# \text{ levels} = 1 + \log_b n$$

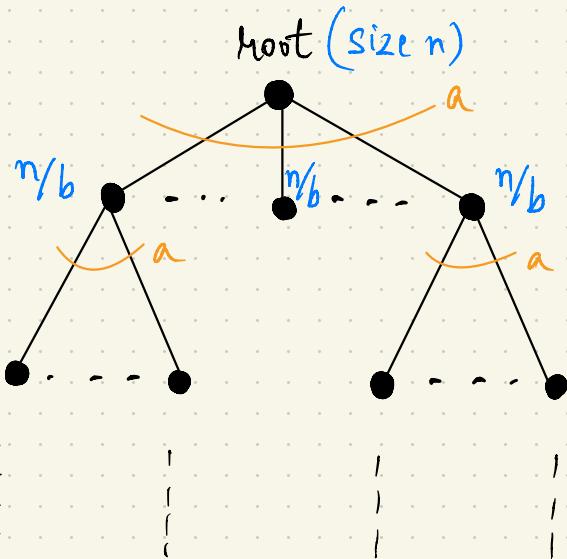
RECURSION TREE



Total work at level j
(ignoring work in recursive calls)

=

RECURSION TREE

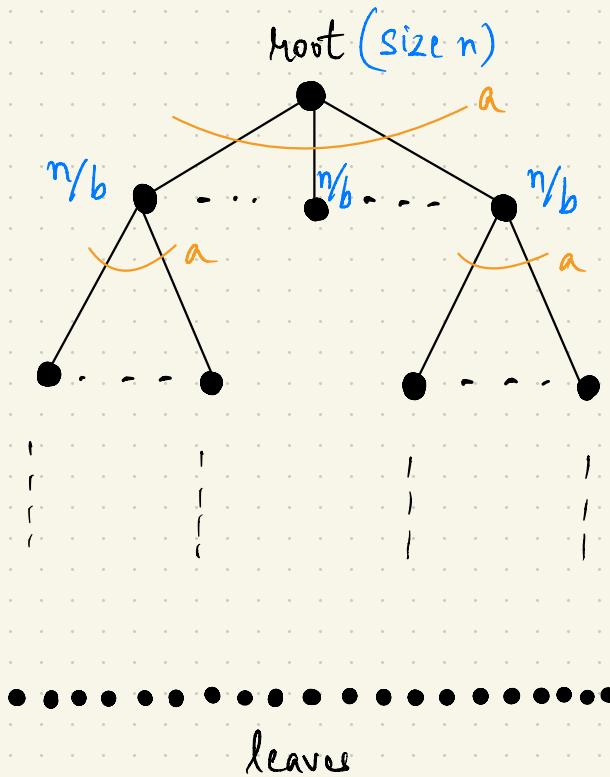


Total work at level j
(ignoring work in recursive calls)

$$= \# \text{ subproblems} \times \text{work done per subproblem}$$

at level j

RECURSION TREE



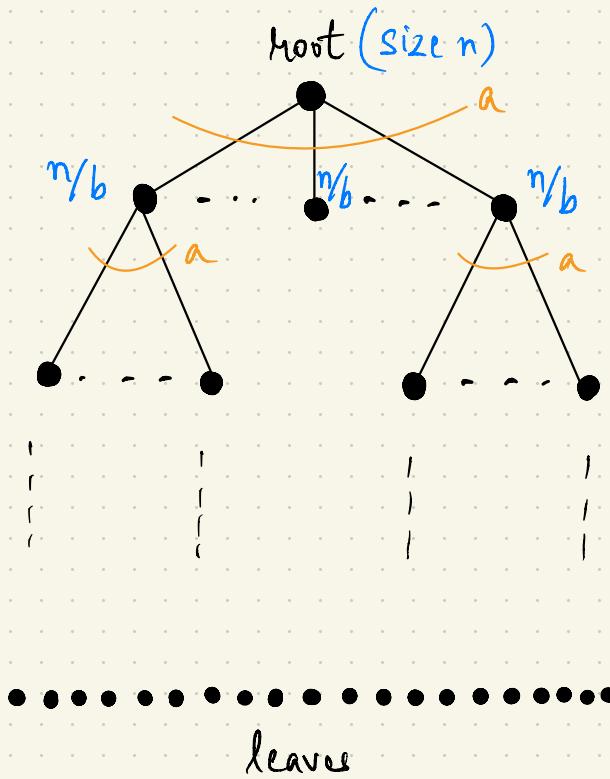
Total work at level j
(ignoring work in recursive calls)

$$= \# \text{ subproblems} \times \text{work done}$$

at level j per subproblem

$$= a^j \times$$

RECURSION TREE



Total work at level j
(ignoring work in recursive calls)

$$= \# \text{ subproblems} \times \text{work done}$$

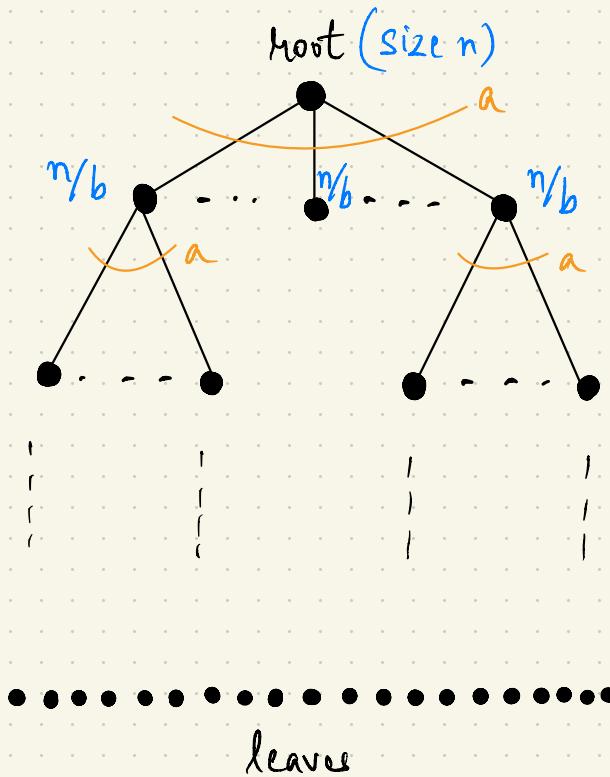
at level j per subproblem

$$= a^j \times$$

Recall general case

$$T(n) \leq a \cdot T\left(\frac{n}{b}\right) + c \cdot n^d$$

RECURSION TREE



Total work at level j
(ignoring work in recursive calls)

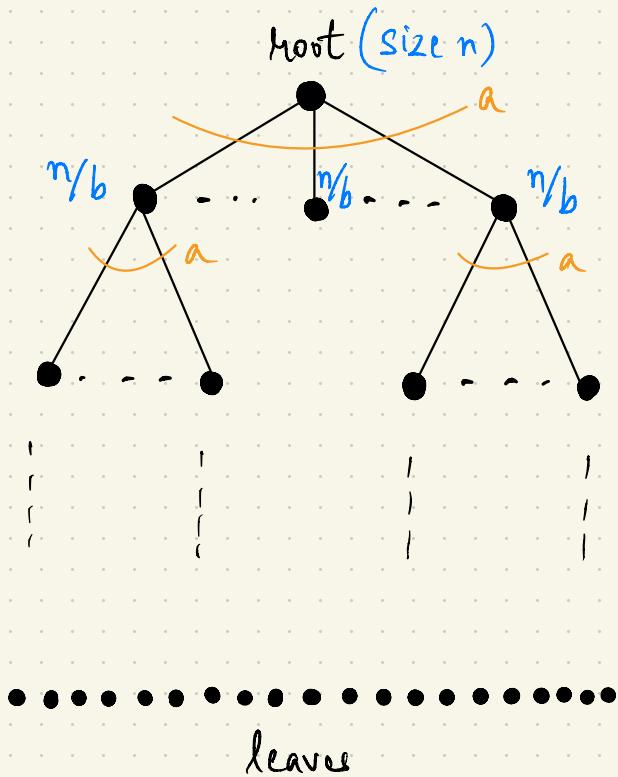
$$= \# \text{ subproblems} \times \text{work done} \\ \text{at level } j \quad \times \text{per subproblem}$$

$$= a^j \times c \cdot \left(\frac{n}{b^j} \right)^d$$

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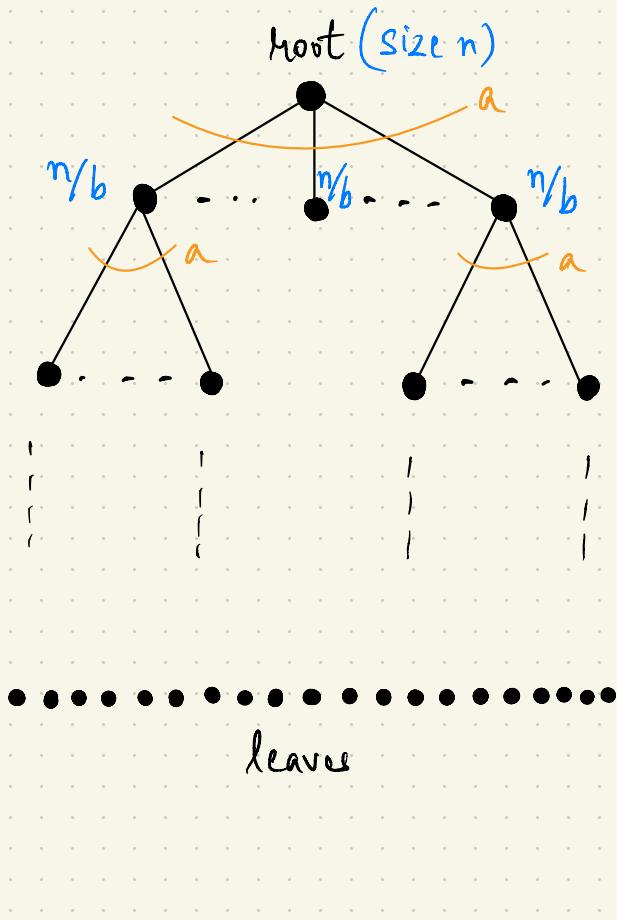
at level j

$$= a^t \times c \cdot \left(\frac{n}{b^t} \right)^d$$

$$= c n^d \cdot \left(\frac{a}{b^d} \right)^j$$

independent of j depends on j

RECURSION TREE



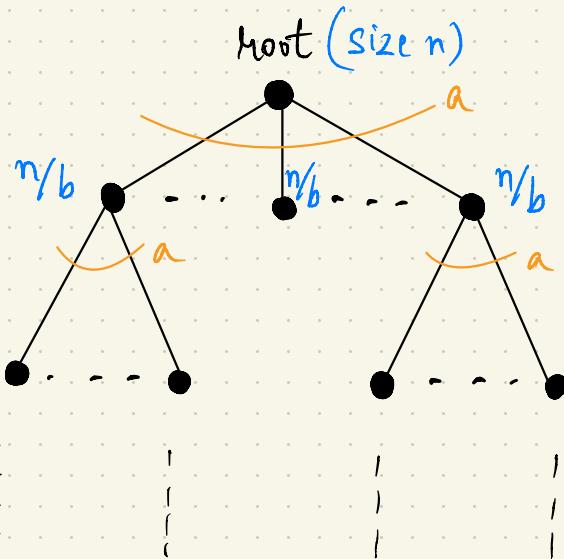
Total work at level j
(ignoring work in recursive calls)

$$= \# \text{ subproblems} \times \text{work done} \\ \text{at level } j \quad \times \text{per subproblem}$$

$$= a^j \times c \cdot \left(\frac{n}{b^j} \right)^d$$

$$= \underbrace{c n^d}_{\text{independent}} \cdot \underbrace{\left(\frac{a}{b^d} \right)^j}_{\text{depends on } j}$$

RECURSION TREE

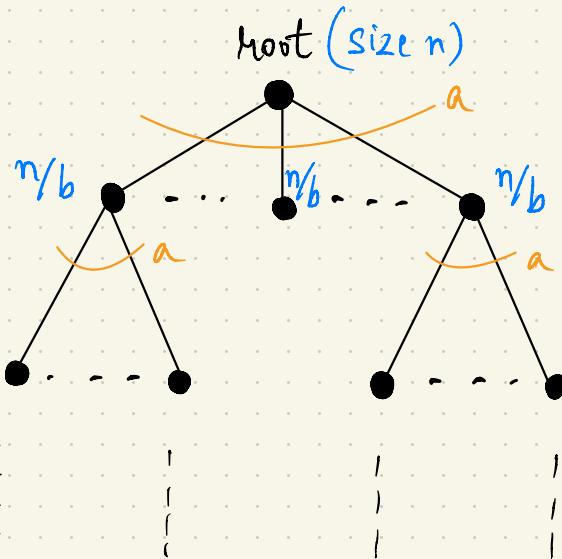


leaves

Total work across all levels

$$\leq c n^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$

RECURSION TREE



leaves

Total work across all levels

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INTERPRETING



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Recall : Work done at level $j = c n^d \cdot \left(\frac{a}{b^d}\right)^j$

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b^d : rate of work shrinkage per subproblem

INTERPRETING 😊

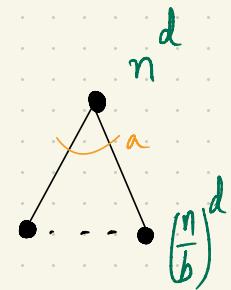
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Recall : Work done at level $j = c n^d \cdot \left(\frac{a}{b^d}\right)^j$

a : rate of problem proliferation

Tug of war between these two opposing forces

b^d : rate of work shrinkage per subproblem

INTERPRETING 😊

Recall : Work done at level $j = cn^d \cdot \left(\frac{a}{b^d}\right)^j$

$$a < b^d$$

$$a > b^d$$

$$a = b^d$$

INTERPRETING 😊

Recall : Work done at level $j = cn^d \cdot \left(\frac{a}{b^d}\right)^j$

$$a < b^d$$

(proliferation < shrinkage)

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(proliferation > shrinkage)

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(proliferation = shrinkage)

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work done per level ↓

most work at root; optimistically $O(n^d)$

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

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most work at the leaves; expect $O(\# \text{leaves})$

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

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work done per level independent of j

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THE CALCULATIONS

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$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



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Case 1 : $a = b^d$

THE CALCULATIONS

$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



Case 1 : $a = b^d$

$$= cn^d (\log_b n + 1)$$

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Case 1 : $a = b^d$

$$= cn^d (\log_b n + 1) = O(n^d \log n)$$

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$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



Case 2 : $a < b^d$

THE CALCULATIONS

$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



Case 2 : $a < b^d$



=

THE CALCULATIONS

$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



Case 2 : $a < b^d$

$$r = a/b^d < 1$$

$$k = \log_b n$$



=

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THE CALCULATIONS

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$$\begin{aligned} &= cn^d (1 + r + r^2 + \dots + r^k) \\ &\leq cn^d (1 + r + r^2 + \dots + r^k + \dots) \\ &= cn^d \frac{1}{1-r} \end{aligned}$$

THE CALCULATIONS

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THE CALCULATIONS

$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



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THE CALCULATIONS

$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



Case 2 : $a < b^d$

$$r = a/b^d < 1$$

dominated by root

$$k = \log_b n$$

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THE CALCULATIONS

$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



Case 3 : $a > b^d$

THE CALCULATIONS

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Case 3 : $a > b^d$

$$r = a/b^d > 1$$

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$$= cn^d \left(\frac{r^{k+1} - 1}{r - 1} \right)$$

$$\leq cn^d \left(\frac{r}{r-1} \right) r^k$$

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$$\text{Total work done} \leq cn^d \cdot \sum_{j=0}^{\log_b n} \left(\frac{a}{b^d}\right)^j$$



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 & \text{Smiley face} = cn^d (1 + r + r^2 + \dots + r^k) \\
 & = cn^d \left(\frac{r^{k+1} - 1}{r - 1} \right) \quad \begin{matrix} \text{\# leaves} \\ \text{!!} \end{matrix} \\
 & \leq cn^d \left(\frac{r}{r-1} \right) r^k = O\left(n^d \left(\frac{a}{b^d}\right)^{\log_b n}\right) = O\left(n^{\log_b a}\right).
 \end{aligned}$$

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Case 3 : $a > b^d$

$$r = a/b^d > 1$$

dominated
by leaves

$$k = \log_b n$$

$$= cn^d (1 + r + r^2 + \dots + r^k)$$

$$= cn^d \left(\frac{r^{k+1} - 1}{r - 1} \right)$$

leaves

!!

$$\leq cn^d \left(\frac{r}{r-1} \right) r^k = O\left(n^d \left(\frac{a}{b^d}\right)^{\log_b n}\right) = O\left(n^{\log_b a}\right).$$

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MASTER THEOREM

Theorem: If $T(n) \leq aT\left(\frac{n}{b}\right) + O(n^d)$, then

$$T(n) = \begin{cases} O(n^d \log n) & \text{if } a = b^d \quad \text{Case 1} \\ O(n^d) & \text{if } a < b^d \quad \text{Case 2} \\ O(n^{\log_b a}) & \text{if } a > b^d \quad \text{Case 3} \end{cases}$$

LIMITATION

Master theorem only applies to equisized subproblems

For more general recurrences , use Akra-Bazzi method.