

Algorithm Analysis, PMI, Asymptotics, and Recurrences

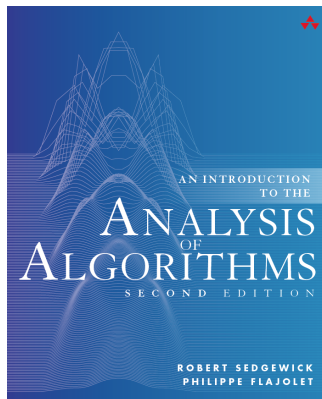
(A Little Mathematics for Computer Science)

Hengfeng Wei

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April 17, 2018





Problem P Algorithm A

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Inputs: \mathcal{X}_n of size n

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Average-case Time Complexity (Problem 1.8)

$$r \in [1, n], r \in \mathbb{Z}^+$$

$$P\{r = i\} = \begin{cases} \frac{1}{n}, & 1 \leq i \leq \frac{n}{4} \\ \frac{2}{n}, & \frac{n}{4} < i \leq \frac{n}{2} \\ \frac{1}{2n}, & \frac{n}{2} < i \leq n \end{cases} \quad T(r) = \begin{cases} 10, & r \leq \frac{n}{4} \\ 20, & \frac{n}{4} < r \leq \frac{n}{2} \\ 30, & \frac{n}{2} < r \leq \frac{3n}{4} \\ n, & \frac{3n}{4} < r \leq n \end{cases}$$

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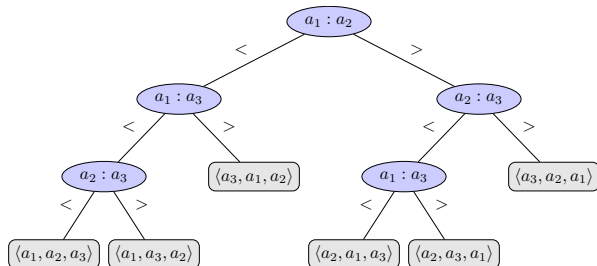
$$\begin{aligned} A &= \sum_{X \in \mathcal{X}} T(X) \cdot P(X) \\ &= T(1)P(1) + T(2)P(2) + \cdots + T(n)P(n) \\ &= \frac{n}{4} \times 10 \times \frac{1}{n} + \frac{n}{4} \times 20 \times \frac{2}{n} + \frac{n}{4} \times 30 \times \frac{1}{2n} + \frac{n}{4} \times n \times \frac{1}{2n} \\ &= \dots \end{aligned}$$

3-element Sorting (Problem 1.1)

- (1) Design an algorithm for **sorting** 3 distinct elements.
- (2) Worst-case and average-case time complexity.
- (3) Worst-case lower bound.

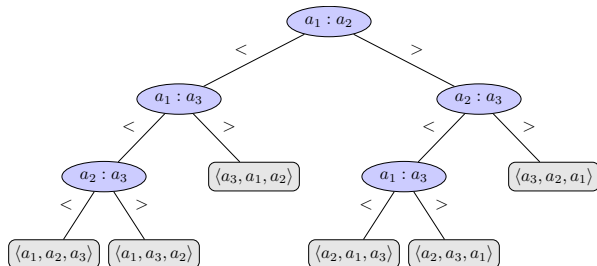
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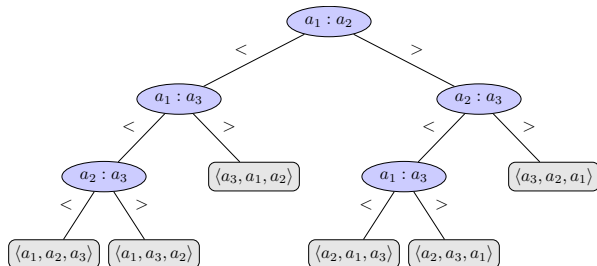
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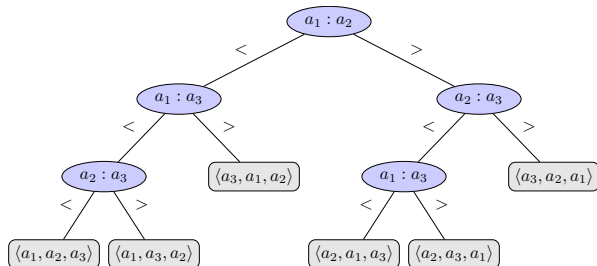
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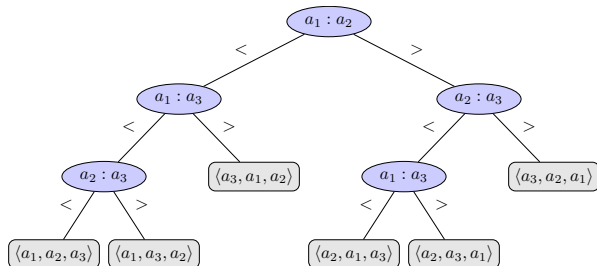
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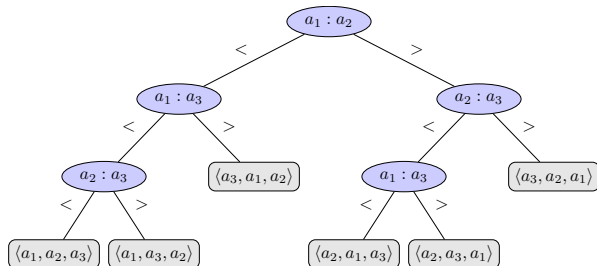
- (1) Design an algorithm for **sorting** 3 distinct elements.
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$$W(3) = 3 \quad B(3) = 2$$

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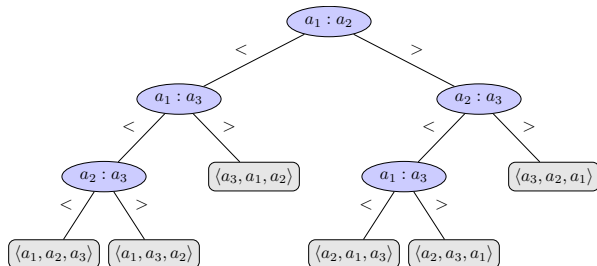
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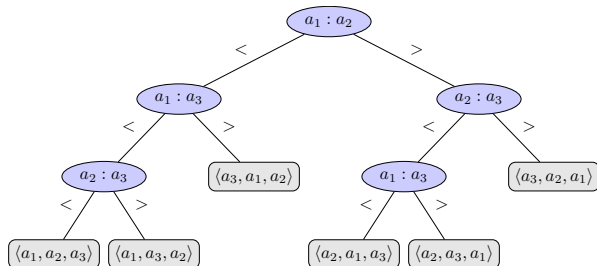
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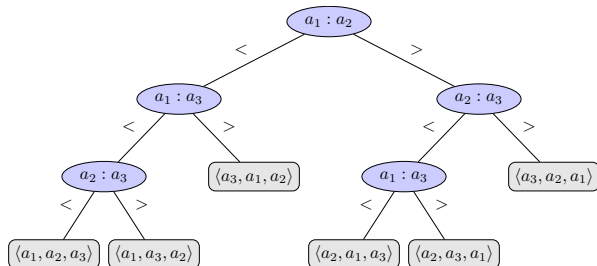


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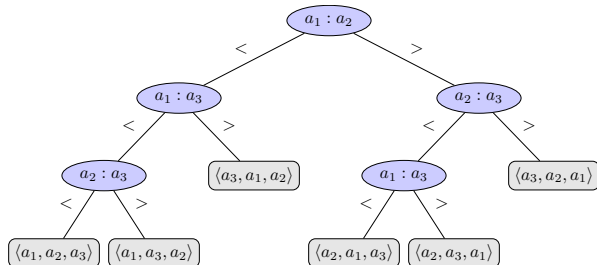


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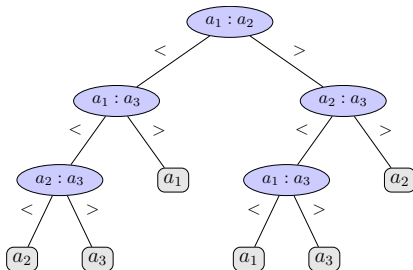
$$LB(3) = 3 \quad (LB(3) \geq \log 3!)$$

3-element Median Selection (Problem 1.2)

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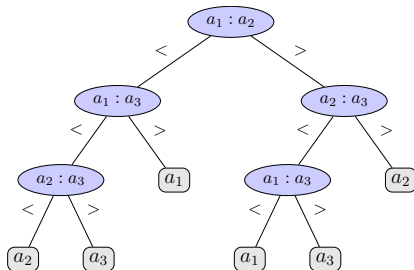
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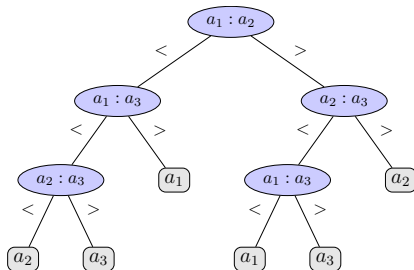
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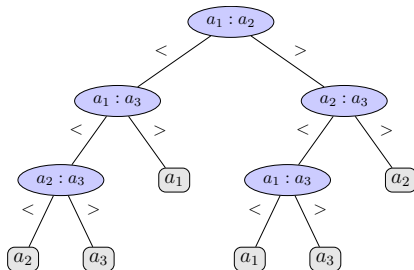


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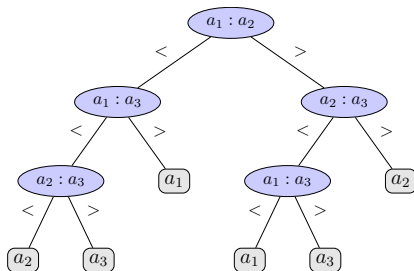


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$$LB(3) = 3 \quad (LB(3) \geq \frac{3n}{2} - \frac{3}{2})$$



$$LB = 2$$



LB = 2

```
1: procedure MEDIAN( $a, b, c$ )
2:   if ( $a - b$ )( $a - c$ ) < 0 then
3:     return  $a$ 
4:   if ( $b - a$ )( $b - c$ ) < 0 then
5:     return  $b$ 
6:   return  $c$ 
```



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Not comparison-based!

Exercise

$$n = 5$$

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Reference

“The Art of Computer Programming, Vol 3: Sorting and Searching (Section 5.3.1)” by Donald E. Knuth

$$S(21) = 66$$

Analysis of Bubblesort (Problem 3.2)

(a) Correctness

(b) $W(n)$ & $A(n)$

(c) Improved version $A'(n)$:

$$l \triangleq \max_k \left\{ \text{SWAP}(A[k], A[k+1]) \right\}$$

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1: procedure BUBBLESORT( $A[1 \cdots n]$ )
2:   for  $i \leftarrow n$  downto 2 do
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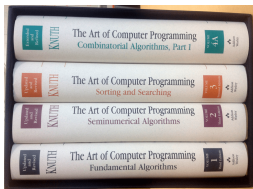
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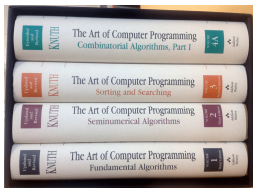
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Then they receive a *practical payoff* when their theories make it possible to get other jobs done more quickly and more economically.

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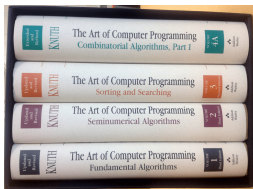


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First of all they experience the sheer *beauty of elegant mathematical patterns* that surround elegant computational procedures.

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Mathematical Induction



Horner's rule (Problem 1.5)

$$P(x) = a_0 + a_1x + a_2x^2 + \cdots + a_{n-1}x^{n-1} + a_nx^n$$

```
1: procedure HORNER( $A[0 \dots n], x$ )                                ▷  $A : \{a_0 \dots a_n\}$ 
2:    $p \leftarrow A[n]$ 
3:   for  $i \leftarrow n - 1$  downto 0 do
4:      $p \leftarrow px + A[i]$ 
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Loop invariant (after the k -th loop):

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When you are in an exam:

20% : Finding \mathcal{I}

80% : Proving \mathcal{I} by PMI

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Base Case: $k = 0$.

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$$\left(\sum_{i=n}^{i=n-k} a_i x^{k-(n-i)} \right) \cdot x + A[n - k - 1] = \sum_{i=n}^{i=n-(k+1)} a_i x^{(k+1)-(n-i)}$$

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Termination

- (a) $i \leftarrow n - 1$ **downto** 0
- (b) $k = n \implies p = \sum_{i=0}^{i=n} a_i x^i$

Integer Multiplication (Problem 1.6)

1: procedure INT-MULT(y, z)	$\triangleright y, z \geq 0; y, z \in \mathbb{Z}$
2: if $z = 0$ then	
3: return 0	
4: return INT-MULT($cy, \lfloor \frac{z}{c} \rfloor$) + $y(z \bmod c)$	$\triangleright c \geq 2$

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I.H.: $< z$ ($z > 0$) : INT-MULT(y, z) = yz .

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4:   return INT-MULT( $cy, \lfloor \frac{z}{c} \rfloor$ ) +  $y(z \bmod c)$       ▷  $c \geq 2$ 
```

Prove by mathematical induction on the non-negative integer z .

BC: $z = 0$: $\text{INT-MULT}(y, 0) = 0 = y \cdot 0$.

I.H.: $z < z$ ($z > 0$) : $\text{INT-MULT}(y, z) = yz$.

I.S.: $= z$.

Integer Multiplication (Problem 1.6)

1: procedure INT-MULT(y, z)	$\triangleright y, z \geq 0; y, z \in \mathbb{Z}$
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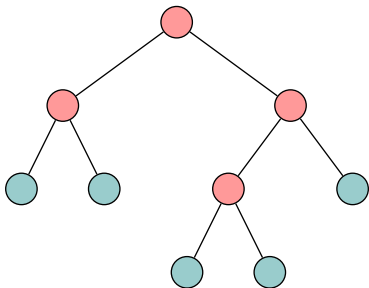
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$$\begin{aligned}\text{INT-MULT}(y, z) &= \text{INT-MULT}(cy, \lfloor \frac{z}{c} \rfloor) + y(z \bmod c) \\ &= cy \cdot \lfloor \frac{z}{c} \rfloor + y(z \bmod c) = yz\end{aligned}$$

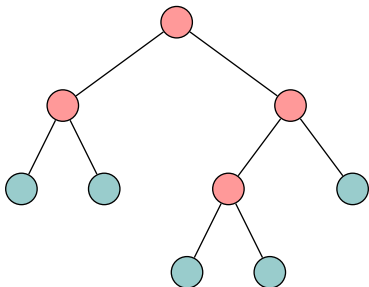
2-tree; Full Binary Tree (Problem 2.5)



$$n_0 = n_2 + 1$$

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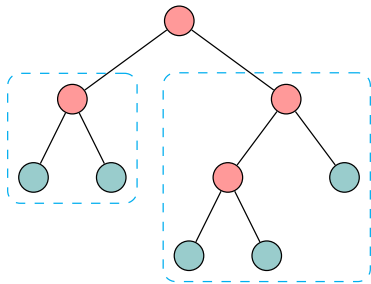
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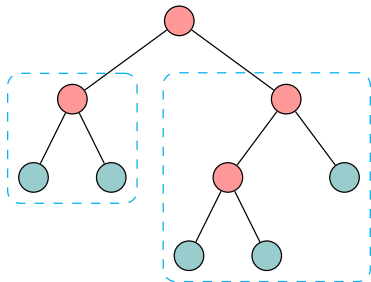
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Prove by mathematical induction on *the size of binary tree*.

2-tree; Full Binary Tree (Problem 2.5)



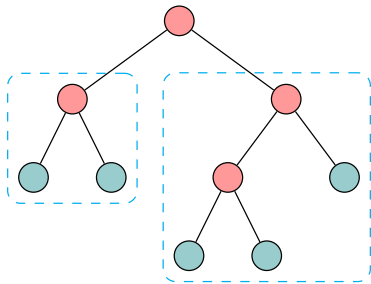
2-tree; Full Binary Tree (Problem 2.5)



$$n_{0L} = n_{2L} + 1$$

$$n_{0R} = n_{2R} + 1$$

2-tree; Full Binary Tree (Problem 2.5)

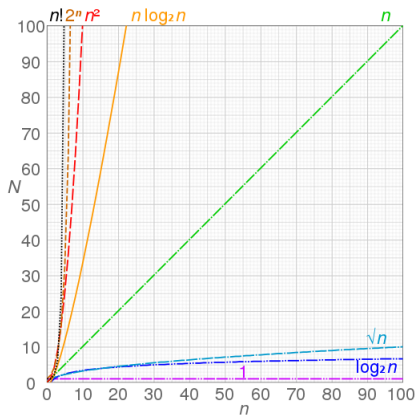


$$n_{0L} = n_{2L} + 1$$

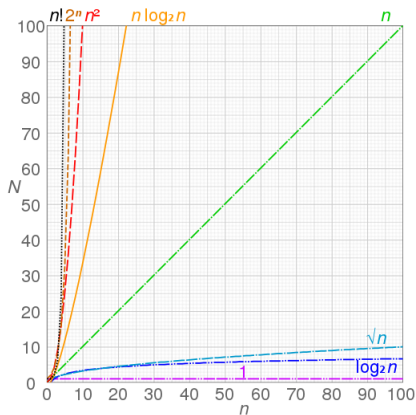
$$n_{0R} = n_{2R} + 1$$

$$n_{0L} + n_{0R} \text{ vs. } n_{2L} + n_{2R} + 1$$

Asymptotics



Asymptotics



$Q : \theta(f) ?$

$$O(g(n)) = \{f(n) \mid \exists c > 0, \exists n_0, \forall n \geq n_0 : 0 \leq f(n) \leq cg(n)\}$$

$$\Omega(g(n)) = \{f(n) \mid \exists c > 0, \exists n_0, \forall n \geq n_0 : 0 \leq cg(n) \leq f(n)\}$$

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$$f(n) \sim g(n) \iff \lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 1$$

Asymptotics (Problem 2.6 (4))

$$f(n) = \Theta(g(n)) \iff f(n) = O(g(n)) \wedge f(n) = \Omega(g(n))$$

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$$f(n) = n, \quad g(n) = n^{1+\sin n}$$

Asymptotics (Problem 2.7 (2))

$$(\log n)^2 \text{ vs. } \sqrt{n}$$

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$$(\log n)^2 \text{ vs. } \sqrt{n}$$

$$(\log n)^{c_1} = O(n^{c_2}) \quad c_1, c_2 > 0$$

Asymptotics (Problem 2.10)

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$$\log(n!) \leq n \log n \qquad \log(n!) \geq \frac{n}{2} \log \frac{n}{2}$$

Summation (Problem 2.20)

```
1: procedure CONUNDRUM( $n$ )
2:    $r \leftarrow 0$ 
3:   for  $i \leftarrow 1$  to  $n$  do
4:     for  $j \leftarrow i + 1$  to  $n$  do
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$$\sum_{i=1}^n \sum_{j=i+1}^n \sum_{k=i+j-1}^n 1 = \frac{1}{48} \left(3(-1 + (-1)^n) + 2n(n+2)(2n-1) \right) = \Theta(n^3)$$

$$\begin{aligned}
& \sum_{i=1}^n \sum_{j=i+1}^n \sum_{k=i+j}^n 1 \\
&= \sum_{i=1}^n \sum_{j=i+1}^n (n - (i+j-1) + 1) [i+j-1 \leq n] \\
&= \sum_{i=1}^n \sum_{j=i+1}^n (n - i - j + 2) [j \leq n - i + 1] \quad n - i + 1 \geq i + 1 \Rightarrow i > \frac{n}{2} \\
&= \sum_{i=1}^{\lfloor \frac{n}{2} \rfloor} \sum_{j=i+1}^{n-i+1} (n - i - j + 2) \\
&= \sum_{i=1}^{\lfloor \frac{n}{2} \rfloor} \sum_{j=1}^{n-2i+1} (n - 2i + 2 - j) \\
& \text{当 } n \text{ 为偶数时} \quad = \frac{1}{2} \sum_{i=1}^{\frac{n}{2}} (n^2 + 3n + 2) + 4 \sum_{i=1}^{\frac{n}{2}} (i^2 - \frac{1}{2}(4n+6)i) \\
&= \frac{1}{2} \times \left(\frac{1}{2}(n^2 + 3n + 2n) + \frac{n(\frac{n}{2}+1)(n+1)}{3} - \frac{(2n+3)(\frac{n}{2}+1)n}{2} \right) \\
&= \frac{2n^3 + 3n^2 - 2n}{24} = \frac{1}{48} (0 + 2n(2+n)(2n-1)) \\
& \text{当 } n \text{ 为奇数时, } \lfloor \frac{n}{2} \rfloor = \frac{n-1}{2}, \text{ 代入, 可化得} = \frac{1}{48} (-6 + 2n(2+n)(2n-1)) \\
& \quad (\text{这个我懒得化了, 谁有兴趣化一下, 多个常数项}) \\
& \text{通解} \quad \frac{1}{48} (3(-1+(-1)^n) + 2n(2+n)(2n-1)) \\
& * \lfloor \frac{n}{2} \rfloor = \frac{n + \frac{(-1)^n + 1}{2}}{2}, \text{ 代入理应可直接得结果, 太繁}
\end{aligned}$$

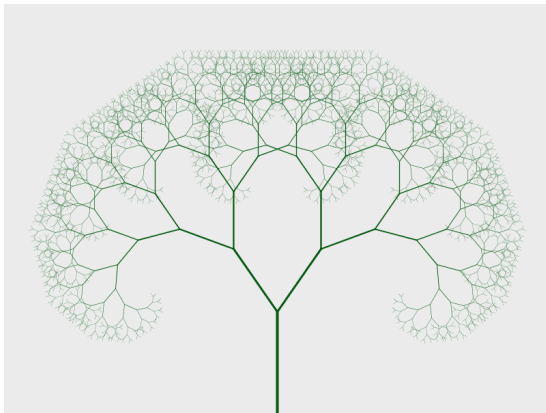
From Zheng (171860658)



Reference:

"Big Omicron and Big Omega and Big Theta" by Donald E. Knuth, 1976.

Recurrences



$$T(n) = aT(n/b) + f(n) \quad (a > 0, b > 1)$$

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$$a^{\log_b n} f(c) = \Theta \left(n^{\log_b a} \begin{array}{l} f(n) \\ af(\frac{n}{b}) \\ a^2 f(\frac{n}{b^2}) \\ \vdots \end{array} \right)$$

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Assume that $T(n)$ is constant for sufficiently small n .

$$\left. \begin{array}{l} f(n) \\ af(\frac{n}{b}) \\ a^2 f(\frac{n}{b^2}) \\ \vdots \\ a^{\log_b n} f(c) = \Theta(n^{\log_b a}) \end{array} \right\} \sum f(n) \stackrel{\text{vs.}}{=} n^E \left\{ \begin{array}{ll} n^{\log_b a} & f(n) = O(n^{E-\epsilon}) \\ n^{\log_b a} \log n & f(n) = \Theta(n^E) \\ f(n) & f(n) = \Omega(n^{E+\epsilon}) \end{array} \right.$$

Solving Recurrences (Problem 2.15)

- (1) $\Theta(n^{\log_3 2})$
- (2) $\Theta(\log^2 n)$
- (3) $\Theta(n)$
- (4) $\Theta(n \log n)$
- (5) $\Theta(n \log^2 n)$
- (6) $\Theta(n^2)$
- (7) $\Theta(n^{\frac{3}{2}} \log n)$
- (8) $\Theta(n)$
- (9) $\Theta(n^{c+1})$
- (10) $\Theta(c^{n+1})$
- (11) \dots

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Gaps in Master Theorem (Problem 2.18)

$$T(n) = 2T(n/2) + \frac{n}{\log n}$$

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Gaps in Master Theorem (Problem 2.18)

$$T(n) = 2T(n/2) + \frac{n}{\log n} = \Theta(n \log \log n)$$

Solving Recurrences (Problem 2.15)

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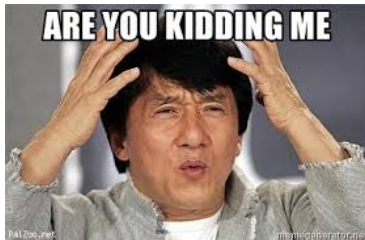
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Solving Recurrences (Problem 2.15 (11))

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$$T(n) = \Theta(n^{0.879146})$$

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$$2^{-\alpha} + 4^{-\alpha} + 8^{-\alpha} = 1$$

Solving Recurrences (Problem 2.15 (11))

$$T(n) = T(n/2) + T(n/4) + T(n/8)$$

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$$T(n) = \Theta(n^\alpha)$$

$$2^{-\alpha} + 4^{-\alpha} + 8^{-\alpha} = 1$$

`Solve[2^{-x} + 4^{-x} + 8^{-x} == 1, x] // N`

Solving Recurrences (Problem 2.15 (11))

$$T(n) = T(n/2) + T(n/4) + T(n/8) + n$$

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Exercise: Prove it by mathematical induction.

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$$T(n) = T(n/2) + T(n/4) + T(n/8) + n$$

By recursion-tree.

$$T(n) = \Theta(n)$$

Exercise: Prove it by mathematical induction.

Reference:

"On the Solution of Linear Recurrence Equations" by Akra & Bazzi, 1996.

$$T(n) = \sum_{i=1}^k a_i T(n/b_i) + f(n)$$

Solving Recurrences (Problem 2.17)

$$\begin{aligned}T(n) &= \sqrt{n} \, T(\sqrt{n}) + n \\&= n^{\frac{1}{2}} \, T\left(n^{\frac{1}{2}}\right) + n \\&= n^{\frac{1}{2}} \left(n^{\frac{1}{2^2}} \, T\left(n^{\frac{1}{2^2}}\right) + n^{\frac{1}{2}} \right) + n \\&= n^{\frac{1}{2} + \frac{1}{2^2}} \, T\left(n^{\frac{1}{2^2}}\right) + 2n \\&= n^{\frac{1}{2} + \frac{1}{2^2}} \left(n^{\frac{1}{2^3}} \, T\left(n^{\frac{1}{2^3}}\right) + n^{\frac{1}{2^2}} \right) + 2n \\&= n^{\frac{1}{2} + \frac{1}{2^2} + \frac{1}{2^3}} \, T\left(n^{\frac{1}{2^3}}\right) + 3n \\&= \dots \\&= n^{\sum_{i=1}^k \frac{1}{2^i}} \, T\left(n^{\frac{1}{2^k}}\right) + kn\end{aligned}$$

$$n^{\frac{1}{2^k}} = 2$$

$$n^{\frac{1}{2^k}} = 2 \implies k = \log \log n$$

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Exercise: Prove it by mathematical induction.

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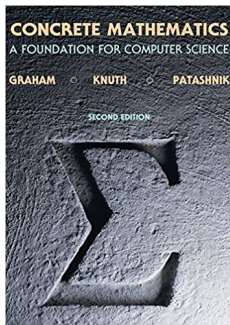
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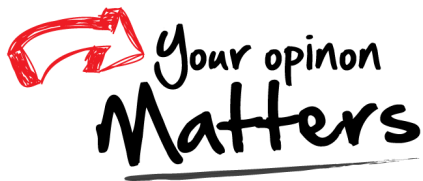
$$T(n) = n \log \log n$$

A Little Mathematics for Computer Science

More Mathematics for Computer Science



Thank
You!



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