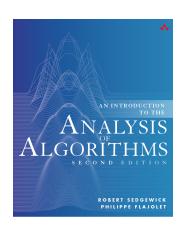
Asymptotics, Recurrences, and Divide and Conquer

Hengfeng Wei

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April 17, 2018





Problem P Algorithm A

$$W(n) = \max_{X \in \mathcal{X}_n} T(X)$$

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Average-case Time Complexity (Problem 1.8)

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

$$P\{r=i\} = \begin{cases} \frac{1}{n}, & 1 \le i \le \frac{n}{4} \\ \frac{2}{n}, & \frac{n}{4} < i \le \frac{n}{2} \\ \frac{1}{2n}, & \frac{n}{2} < i \le n \end{cases}$$

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$$A = \sum_{X \in \mathcal{X}} T(X) \cdot P(X)$$

$$= T(1)P(1) + T(2)P(2) + \dots + T(n)P(n)$$

$$= \frac{n}{4} \times 10 \times \frac{1}{n} + \frac{n}{4} \times 20 \times \frac{2}{n} + \frac{n}{4} \times 30 \times \frac{1}{2n} + \frac{n}{4} \times n \times \frac{1}{2n}$$

$$= \frac{1}{8}n + \frac{65}{4}$$

Average-case Analysis of Quicksort

$$A(n) = n - 1 + \frac{1}{n} \sum_{i=0}^{i=n-1} (A(i) + A(n-i-1))$$
$$A(n) = \underset{X \in \mathcal{X}_n}{\mathbb{E}} [T(X)] = \sum_{X \in \mathcal{X}_n} T(X) \cdot P(X)$$

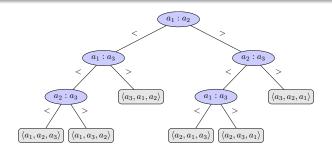
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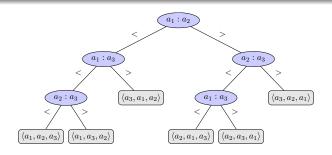
$$\begin{split} A(n) &= \mathbb{E}[T(X)] \\ &= \mathbb{E}[\mathbb{E}[T(X)|I]] \\ &= \sum_{i=0}^{i=n-1} P(I=i) \; \mathbb{E}[T(X) \mid I=i] \\ &= \sum_{i=0}^{i=n-1} \frac{1}{n}[n-1+A(i)+A(n-i-1)] \end{split}$$

- 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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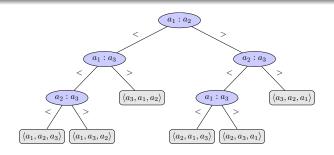


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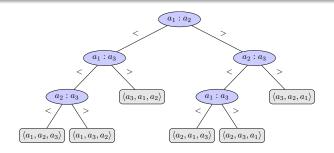
$$W(3) =$$

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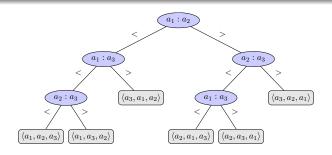
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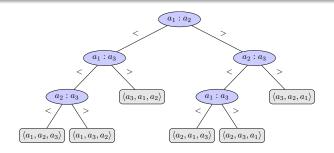
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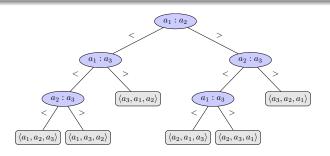
$$W(3) = 3$$
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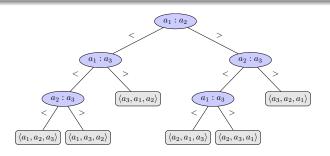
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 $B(3) = 2$ $A(3) = \frac{1}{6}(3+3+2+3+3+2) = \frac{8}{3}$

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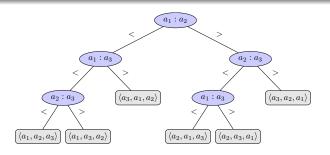


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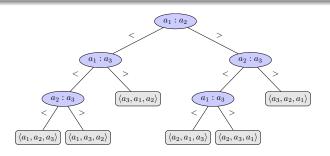


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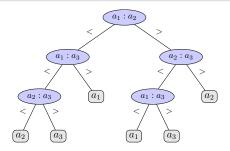


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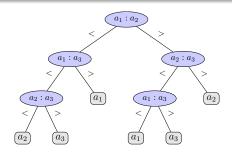
$$\mathsf{LB}(3) = 3 \qquad (\mathsf{LB}(3) \ge \log 3!)$$

- 3-element Median Seletion (Problem 1.2)
- (1) Design an algorithm for selecting the median of 3 distinct elements.
- (2) Worst-case and average-case time complexity.
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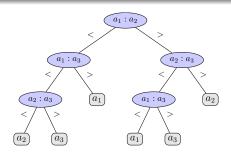


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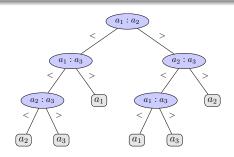


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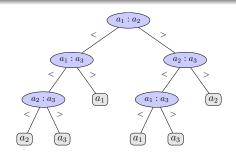


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$$W(3) = 3$$
 $B(3) = 2$ $A(3) = \frac{8}{3}$
 $\mathsf{LB}(3) = 3$ $(\mathsf{LB}(3) \ge \frac{3n}{2} - \frac{3}{2})$



LB = 2



$$LB = 2$$

```
1: procedure \operatorname{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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LB = 2

Not comparison-based!

Exercise

$$n = 5$$

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Exercise

$$n=5$$

Reference

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

$$S(21) = 66$$



Mathematical Induction



Horner's rule (Problem 1.5)

$$P(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{n-1} x^{n-1} + a_n x^n$$

1: **procedure** HORNER(A[0...n], x)

 $\triangleright A:\{a_0\ldots a_n\}$

- 2: $p \leftarrow A[n]$
- 3: for $i \leftarrow n-1$ downto 0 do
- 4: $p \leftarrow px + A[i]$
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When you are in an exam:

20%: Finding \mathcal{I}

80%: Proving $\mathcal I$ by PMI

$$\mathcal{I}: p = \sum_{i=n}^{i=n-k} a_i x^{k-(n-i)}$$

Proof.

Prove by mathematical induction on non-negative integer k, the number of loops.

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Proof.

Prove by mathematical induction on non-negative integer k, the number of loops.

Basis:

$$k=0: p=a_n=\mathcal{I}_0$$

Inductive Hypothesis:

Inductive Step:



Integer Multiplication (Problem 1.6)

- 1: **procedure** Int-Mult(y, z)
- 2: if z = 0 then
- 3: return 0
- 4: **return** INT-MULT $(cy, \lfloor \frac{z}{c} \rfloor) + y(z \mod c)$

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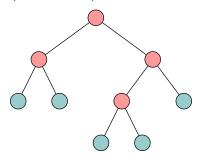
Proof.

Prove by mathematical induction on non-negative integer z.





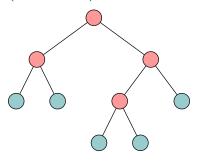
2-tree; full binary tree (Problem 2.5)



$$n_0 = n_2 + 1$$

Proof.

2-tree; full binary tree (Problem 2.5)



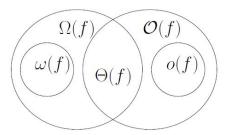
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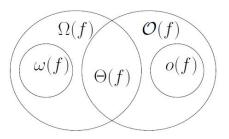
Prove by mathematical induction on the structure of binary tree.



Asymptotics



Asymptotics



$$Q:\theta(f)$$
?



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

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

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

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

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Asymptotics (Problem 2.6(6))

$$\Theta(g(n))\cap o(g(n))=\emptyset$$

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$$Q: f(n) = O(g(n)) \lor g(n) = \Omega(f(n))?$$

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

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Reference:

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

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

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

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

$$\log(n!) = \Theta(n \log n)$$

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$$\log(n!) = \log 1 + \log 2 + \dots + \log n$$

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$$\log(n!) = \log 1 + \log 2 + \dots + \log n$$

$$\log(n!) \le n \log n \qquad \log(n!) \ge \frac{n}{2} \log \frac{n}{2}$$



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
5: for k \leftarrow i+j-1 to n do
6: r \leftarrow r+1
7: return r
```

Summation (Problem 2.20)

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$$\sum_{i=1}^{n} \sum_{j=i+1}^{n} \sum_{k=i+j-1}^{n} 1 =$$



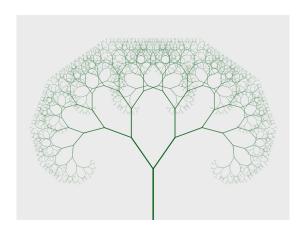
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$$\sum_{i=1}^{n} \sum_{j=i+1}^{n} \sum_{k=i+j-1}^{n} 1 = \frac{n^2 - n}{2} = \Theta(n^2)$$



Recurrences



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

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

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$$\begin{cases} f(n) \\ af(\frac{n}{b}) \\ a^2f(\frac{n}{b^2}) \\ \vdots \\ a^{\log_b n}f(c) = \Theta(n^{\log_b a}) \end{cases} \sum_{\substack{f(n) \text{ vs. } n^E \\ =}} \begin{cases} n^{\log_b a} & f(n) = O(n^{E-\epsilon}) \\ n^{\log_b a} \log n & f(n) = \Theta(n^E) \\ f(n) & f(n) = \Omega(n^{E+\epsilon}) \end{cases}$$

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) $\Theta(c^{n+1})$
- $(11) \cdots$

$$T(n) = T(n/2) + \log n$$
$$T(n) = 2T(n/2) + n \log n$$

Solving Recurrences

(Problem 2.15)

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

$$T(n) = T(n/2) + \log n$$
$$T(n) = 2T(n/2) + n \log n$$

Reference:

$$\underline{f(n) = \Theta(n^{\log_b a} \log^k n) \implies \Theta(n^{\log_b a} \log^{k+1} n)}$$



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)
$$\Theta(c^{n+1})$$

 $(11) \cdots$

$$T(n) = T(n/2) + \log n$$

$$T(n) = 2T(n/2) + n \log n$$

Reference:

$$f(n) = \Theta(n^{\log_b a} \log^k n) \implies \Theta(n^{\log_b a} \log^{k+1} n)$$

$$T(n) = T(n-1) + n^c \quad c \ge 1$$

$$T(n) = T(n-1) + c^n \quad c > 1$$

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

$$T(n) = T(n/2) + T(n/4) + T(n/8) + \frac{n}{n}$$

$$T(n) = T(n/2) + T(n/4) + T(n/8) + \frac{n}{2}$$

By recursion-tree.

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

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By recursion-tree.

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

Exercise: Prove it by mathematical induction.

$$T(n) = T(n/2) + T(n/4) + T(n/8) + \frac{n}{2}$$

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.

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

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

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

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

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

$$T(n) = n^{\sum_{i=1}^{k} \frac{1}{2^{i}}} T\left(n^{\frac{1}{2^{k}}}\right) + kn$$
$$= n^{\sum_{i=1}^{\log \log n} \frac{1}{2^{i}}} T(2) + n \log \log n$$

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

$$\mathsf{T}(n) = n^{\sum_{i=1}^{k} \frac{1}{2^i}} \mathsf{T}\left(n^{\frac{1}{2^k}}\right) + kn$$
$$= n^{\sum_{i=1}^{\log \log n} \frac{1}{2^i}} \mathsf{T}(2) + n \log \log n$$

$$\sum_{i=1}^{\log_2 \log_2(n)} \frac{1}{2^i} < 1 \implies T(n) = \Theta(n \log \log n)$$

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

$$\mathsf{T}(n) = n^{\sum_{i=1}^{k} \frac{1}{2^i}} \mathsf{T}\left(n^{\frac{1}{2^k}}\right) + kn$$
$$= n^{\sum_{i=1}^{\log \log n} \frac{1}{2^i}} \mathsf{T}(2) + n \log \log n$$

$$\sum_{i=1}^{\log_2\log_2(n)}\frac{1}{2^i}<1 \implies T(n)=\Theta(n\log\log n)$$

Exercise: Prove it by mathematical induction.



$$T(n) = \sqrt{n}T(\sqrt{n}) + n$$

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$$S(m) = S(m/2) + 1 =$$

$$T(n) = \sqrt{n}T(\sqrt{n}) + n$$

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$$S(m) = S(m/2) + 1 = \Theta(\log m)$$

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$$S(m) = S(m/2) + 1 = \Theta(\log m)$$

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

