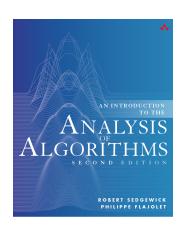
Asymptotics, Recurrences, and Divide and Conquer

Hengfeng Wei

hfwei@nju.edu.cn

April 17, 2018





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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

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

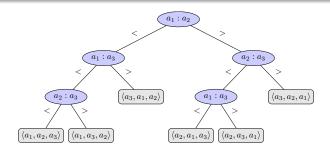
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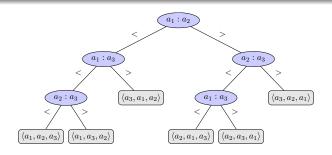
$$\begin{split} A(n) &= \mathbb{E}[T(X)] \\ &= \mathbb{E}[\mathbb{E}[T(X)|I]] \\ &= \sum_{i=0}^{i=n-1} P(I=i) \; \mathbb{E}[T(X) \mid I=i] \\ &= \sum_{i=0}^{i=n-1} \frac{1}{n}[n-1+A(i)+A(n-i-1)] \end{split}$$

- 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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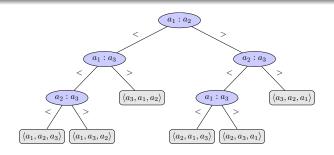


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$$W(3) =$$

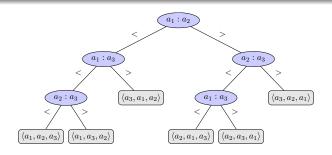
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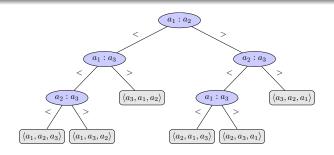


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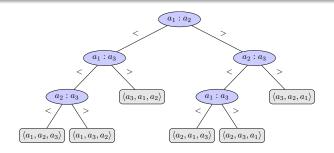
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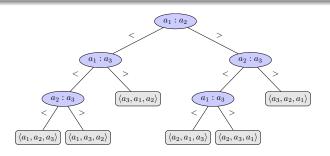
$$W(3) = 3$$
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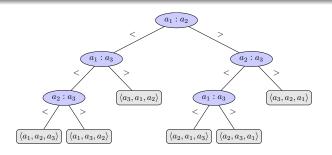
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 $B(3) = 2$ $A(3) = \frac{1}{6}(3+3+2+3+3+2) = \frac{8}{3}$

↓□▶ ↓□▶ ↓□▶ ↓□▶ ↓□ ♥ ♀○

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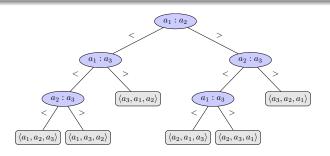


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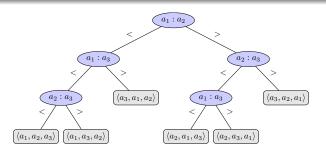


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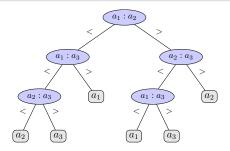


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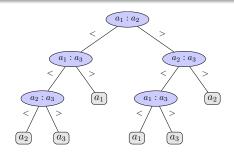
$$\mathsf{LB}(3) = 3 \qquad (\mathsf{LB}(3) \ge \log 3!)$$

- 3-element Median Seletion (Problem 1.2)
- (1) Design an algorithm for selecting the median of 3 distinct elements.
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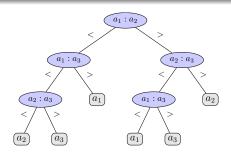


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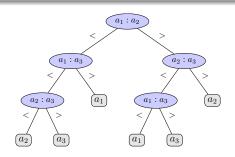


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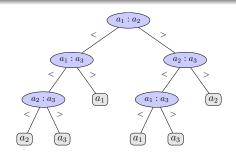


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$$LB(3) = 3$$



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$$W(3) = 3 \qquad B(3) = 2 \qquad A(3) = \frac{8}{3}$$

$$\mathsf{LB}(3) = 3 \qquad (\mathsf{LB}(3) \ge \frac{3n}{2} - \frac{3}{2})$$



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

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6: return c
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LB = 2

Not comparison-based!

Exercise

$$n = 5$$

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.

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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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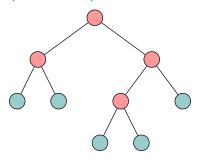
Proof.

Prove by mathematical induction on non-negative integer z.





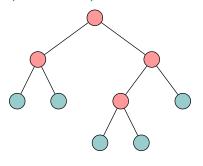
2-tree; full binary tree (Problem 2.5)



$$n_0 = n_2 + 1$$

Proof.

2-tree; full binary tree (Problem 2.5)

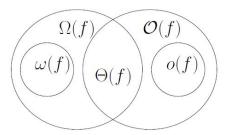


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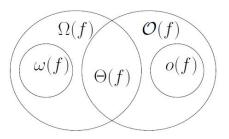
Proof.

Prove by mathematical induction on the structure of binary tree.

Asymptotics



Asymptotics



 $Q:\theta(f)$?

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

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

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

