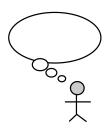
2-2 The Efficiency of Algorithms

魏恒峰

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2018年04月02日

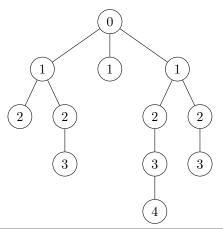
- (1) Diameter of Convex Polygon: $\Theta(n)$
- (2) Lower Bound for Sorting: $\Omega(n \lg n)$
- (3) Traversal over Trees: DFS/BFS $(\Theta(n))$



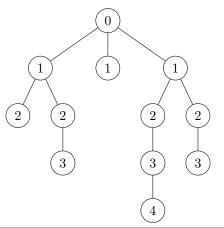
- (1) Diameter of Convex Polygon: $\Theta(n)$
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- (3) Traversal over Trees: DFS/BFS $(\Theta(n))$

I have thought that · · ·

DH 4.2 (a): Sum of Depths



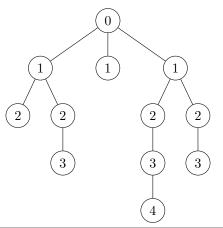
DH 4.2 (a): Sum of Depths



$$\mathsf{sum-of-depths}(r) = \left\{ \begin{array}{l} \sum\limits_{v: \mathsf{child of}\ r} \mathsf{sum-of-depths}(v) + \mathsf{depth of}\ r, \end{array} \right.$$

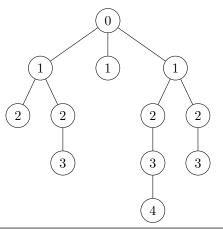
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DH 4.2 (a): Sum of Depths



$$\mathsf{sum\text{-}of\text{-}depths}(r) = \left\{ \begin{array}{ll} \mathsf{depth} \ \mathsf{of} \ r, & r \ \mathsf{is} \ \mathsf{a} \ \mathsf{leaf} \\ \sum\limits_{v: \mathsf{child} \ \mathsf{of} \ r} \mathsf{sum\text{-}of\text{-}depths}(v) + \mathsf{depth} \ \mathsf{of} \ r, & \mathsf{o.w.} \end{array} \right.$$

DH 4.2 (a): Sum of Depths

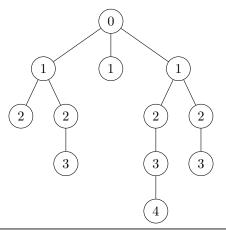


$$\mathsf{sum-of-depths}(r, \textcolor{red}{d}) = \left\{ \begin{array}{l} d, & r \text{ is a leaf} \\ \sum\limits_{v: \mathsf{child of } r} \mathsf{sum-of-depths}(v, \textcolor{red}{d} + 1) + d, & \mathsf{o.w.} \end{array} \right.$$

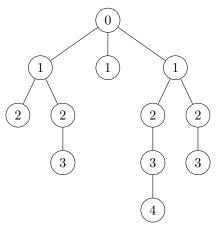
Algorithm 1 Calculate the sum of depths of all nodes of a tree T.

- 1: procedure Sum-of-Depths()
- 2: **return** SUM-OF-DEPTHS(T, 0)
- 3: **procedure** Sum-of-Depths(r, depth) $\triangleright r$: root of a tree
- 4: **if** T is a leaf **then**
- 5: **return** depth
- 6: **for all** child vertex v of r **do**
- 7: $depth \leftarrow depth + \text{Sum-of-Depths}(v, depth + 1)$
- 8: **return** depth

DH 4.2 (b): Number of Nodes at Depth K



DH 4.2 (b): Number of Nodes at Depth K

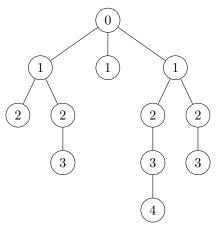


$$\mathsf{nodes}\text{-}\mathsf{at}\text{-}\mathsf{depth}(r, \pmb{k}) =$$

 $\mathsf{nodes}\text{-}\mathsf{at}\text{-}\mathsf{depth}(v, \frac{k}{k} - 1),$

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DH 4.2 (b): Number of Nodes at Depth K



$$\mathsf{nodes-at-depth}(r, \pmb{k}) = \left\{ \begin{array}{l} 1, & k = 0 \\ 0, & k > 0 \wedge r \text{ is a leaf} \\ \sum & \mathsf{nodes-at-depth}(v, \pmb{k-1}), & \mathsf{o.w.} \end{array} \right.$$

v:child of r2-2 The Efficiency of Algorithms

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Algorithm 2 Count the number of nodes in T at depth K.

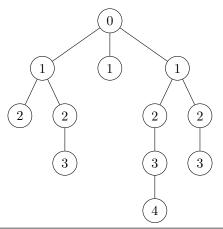
- 1: procedure Nodes-At-Depth()
- 2: **return** Nodes-At-Depth(T, K)

```
3: procedure Nodes-At-Depth(r, k)
```

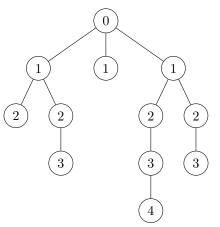
- 4: if k = 0 then
- 5: return 1
- 6: **if** r is a leaf **then**
- 7: **return** 0
- 8: $num \leftarrow 0$
- 9: **for all** child vertex v of r **do**
- 10: $num \leftarrow num + \text{Nodes-at-Depth}(v, k 1)$
- 11: return num

 $\triangleright r$: root of a tree

DH 4.2 (c): Any Leaf at an Even Depth?



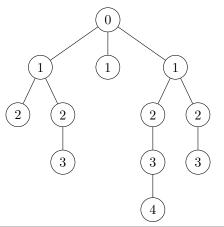
DH 4.2 (c): Any Leaf at an Even Depth?



$$\mathsf{leaf-at-depth}(r, \underbrace{parity}) = \left\{ \begin{array}{l} \sum\limits_{v: \mathsf{child of } r} (v, \mathbf{1} - parity), \end{array} \right.$$

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DH 4.2 (c): Any Leaf at an Even Depth?



$$\mathsf{leaf-at-depth}(r, \underbrace{parity}) = \left\{ \begin{array}{ll} 1 - parity, & r \text{ is a leaf} \\ \sum\limits_{v: \mathsf{child of } r} (v, 1 - parity), & \mathsf{o.w.} \end{array} \right.$$

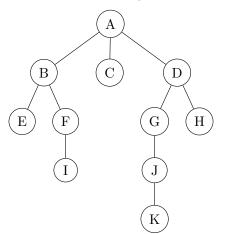
9 / 32

Algorithm 3 Check whether a tree T has any leaf at an even depth.

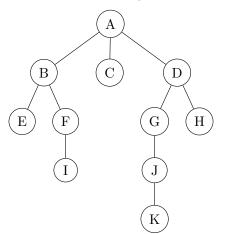
- 1: procedure Leaf-at-Even-Depth()
- 2: **return** Leaf-at-Depth(T, even = 0)
- 3: **procedure** Leaf-at-Depth(r, parity)
- 4: **if** r is a leaf **then**
- 5: **return** 1 parity
- 6: $result \leftarrow 0$
- 7: **for all** child vertex v of r **do**
- 8: $result \leftarrow result \lor \text{Leaf-at-Depth}(v, 1 parity)$
- 9: return result

 $\triangleright r$: root of a tree

DH 4.3 (a): Sum of Contents at Each Depth



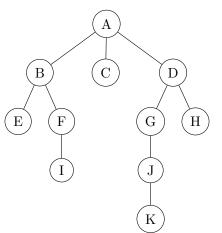
DH 4.3 (a): Sum of Contents at Each Depth



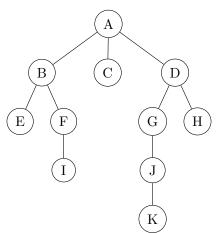
Algorithm 4 Calculate the sum of contents of nodes of a tree T at each depth.

```
1: procedure SUM-AT-DEPTH(r)
                                                               \triangleright r: root of the tree T
        r.depth \leftarrow 0
 2:
        Q \leftarrow \emptyset
 3:
         ENQUEUE(Q, r)
 4:
        while Q \neq \emptyset do
 5:
             u \leftarrow \text{DEQUEUE}(Q)
 6:
             sumAtDepth[u.depth] += u.content
 7:
             for all child vertex v of u do
 8.
                 v.depth \leftarrow u.depth + 1
 9.
                 ENQUEUE(Q, v)
10:
```

DH 4.3 (b): Depth K with the Maximum Number of Nodes



DH 4.3 (b): Depth K with the Maximum Number of Nodes



Algorithm 5 Count the number of nodes of a tree T at each depth.

```
1: procedure Nodes-At-Depth(r)
                                                               \triangleright r: root of the tree T
        r.depth \leftarrow 0
 2:
        Q \leftarrow \emptyset
 3:
        ENQUEUE(Q, r)
 4:
        while Q \neq \emptyset do
 5:
             u \leftarrow \text{DEQUEUE}(Q)
 6:
             nodesAtDepth[u.depth] += 1
 7:
             for all child vertex v of u do
 8:
                 v.depth \leftarrow u.depth + 1
 9:
                  ENQUEUE(Q, v)
10:
```

Lower Bound for Comparion-based Sorting

Lower Bound for Comparion-based Sorting



Prove a lower bound of $O(n \lg n)$ on the time complexity of any comparison-based sorting algorithm.

Prove a lower bound of $O(n\lg n)$ on the time complexity of any comparison-based sorting algorithm.



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Lower Bound for Comparison-based Sorting (DH 6.13)

Prove a lower bound of $\Omega(n \lg n)$ on the time complexity of any comparison-based sorting algorithm.

Prove a lower bound of $O(n\lg n)$ on the time complexity of any comparison-based sorting algorithm.



Lower Bound for Comparison-based Sorting (DH 6.13)

Prove a lower bound of $\Omega(n \lg n)$ on the time complexity of any comparison-based sorting algorithm on inputs of size n.

Prove a lower bound of $\Omega(n \lg n)$ on the time complexity of any **comparison-based** sorting algorithm on inputs of size n.

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Computational Model:

the only way to gain order info.

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Computational Model:

the only way to gain order info.

$$x \in [1 \cdots 99]$$
$$x/10$$

Prove a lower bound of $\Omega(n \lg n)$ on the time complexity of any **comparison-based** sorting algorithm on inputs of size n.

Cost Model:

Computational Model:

the critical operations to count

the only way to gain order info.

$$x \in [1 \cdots 99]$$
$$x/10$$

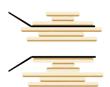
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Cost Model:

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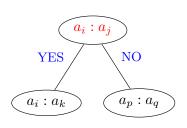


the critical operations to count

"Bounds For Sorting By Prefix Reversal", 1979

Decision Tree Model

Decision Tree Model



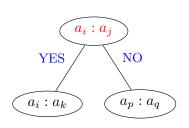
Nodes: comparisions $a_i : a_j$

$$a_i < a_j, \ a_i \le a_j, \ a_i = a_j$$

$$a_i \ge a_j, \ a_i > a_j$$

Edges: two-way decisions

Leaves: possible permutations



Nodes: comparisions $a_i : a_j$

$$a_i < a_j, \ a_i \le a_j, \ a_i = a_j$$

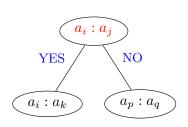
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Assumption (By aware of any assumptions !!!):

All the input elements are distinct.



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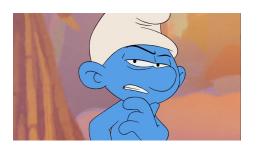
All the input elements are **distinct**.

$$a_i < a_j$$

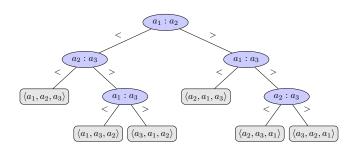


Any Comparison-based Sorting Algorithm $\xrightarrow{\text{modeled by}}$ A Decision Tree

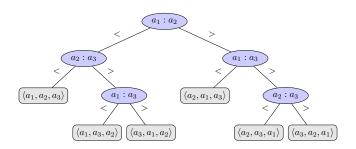
Any Comparison-based Sorting Algorithm $\xrightarrow{\text{modeled by}}$ A Decision Tree



Any Comparison-based Sorting Algorithm $\stackrel{\text{modeled by}}{\longrightarrow}$ A Decision Tree



Any Comparison-based Sorting Algorithm $\stackrel{\text{modeled by}}{\longrightarrow}$ A Decision Tree



The decision tree for insertion sort on three elements.

Any Comparison-based Sorting Algorithm $\xrightarrow{\text{modeled by}}$ A Decision Tree

Any Comparison-based Sorting Algorithm $\stackrel{\text{modeled by}}{\longrightarrow}$ A Decision Tree

```
1: procedure -SORT(A, n)

2: for i \leftarrow 1 to n-1 do

3: for j \leftarrow i+1 to n do

4: if A[j] < A[i] then

5: SWAP(A[j], A[i])
```

Any Comparison-based Sorting Algorithm $\stackrel{\text{modeled by}}{\longrightarrow}$ A Decision Tree

```
1: procedure SELECTION-SORT(A, n)

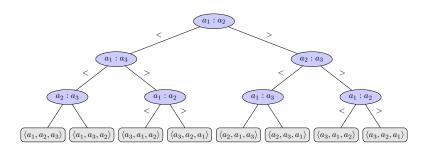
2: for i \leftarrow 1 to n-1 do

3: for j \leftarrow i+1 to n do

4: if A[j] < A[i] then

5: SWAP(A[j], A[i])
```

Any Comparison-based Sorting Algorithm $\xrightarrow{\text{modeled by}}$ A Decision Tree



The decision tree for selection sort on three elements.

Any Comparison-based Sorting Algorithm $\mathcal{A} \xrightarrow{\mathsf{modeled}\ \mathsf{by}} \mathsf{A}$ Decision Tree \mathcal{T}

Any Comparison-based Sorting Algorithm $\mathcal{A} \xrightarrow{\mathsf{modeled}\ \mathsf{by}} \mathsf{A}$ Decision Tree \mathcal{T}

Algorithm ${\mathcal A}$ on a specific input of size $n \xrightarrow{\mathsf{modeled}\ \mathsf{by}} \mathsf{A}$ path through ${\mathcal T}$

Any Comparison-based Sorting Algorithm $\mathcal{A} \xrightarrow{\mathsf{modeled}\ \mathsf{by}} \mathsf{A}$ Decision Tree \mathcal{T}

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Worst-case time complexity of $\mathcal{A} \xrightarrow{\mathsf{modeled}\ \mathsf{by}}$ The height of \mathcal{T}

Any Comparison-based Sorting Algorithm $\mathcal{A} \xrightarrow{\mathsf{modeled}\ \mathsf{by}} \mathsf{A}$ Decision Tree \mathcal{T}

Algorithm $\mathcal A$ on a specific input of size $n \xrightarrow{\mathsf{modeled}\ \mathsf{by}} \mathsf A$ path through $\mathcal T$

Worst-case time complexity of $\mathcal{A} \stackrel{\mathsf{modeled\ by}}{\longrightarrow}$ The height of \mathcal{T}

Worst-case Lower Bound of Comparison-based Sorting on inputs of size n $\underline{ \text{modeled by} }$

The Minimum Height of All \mathcal{T} s

The Minimum Height of All $\mathcal{T}s$

The Minimum Height of All \mathcal{T} s

To be a full binary tree:

$$\#$$
 of leaves $\leq 2^h$

The Minimum Height of All \mathcal{T} s

To be a full binary tree:

$$\#$$
 of leaves $\leq 2^h$

To be a correct sorting algorithm:

$$\#$$
 of leaves $> n!$



Lower Bound for Comparison-based Sorting

 $n! \le \#$ of leaves $\le 2^h$

Lower Bound for Comparison-based Sorting

$$n! \le \#$$
 of leaves $\le 2^h$

$$h \ge \lg n! = \Omega(n \lg n)$$

Lower Bound for Comparison-based Sorting

$$n! \leq \# \text{ of leaves} \leq 2^h$$

$$h \ge \lg n! = \Omega(n \lg n)$$

Stirling Formula (by James Stirling):

$$n! = \Theta(\sqrt{2\pi n} \left(\frac{n}{e}\right)^n)$$

Proof.



The k-sorted Problem

An array $A[1\cdots n]$ is k-sorted if it can be divided into k blocks, each of size n/k (we assume that $n/k\in\mathbb{N}$), such that the elements in each block are larger than the elements in earlier blocks and smaller than elements in later blocks. The elements within each block need not be sorted.

- (a) Describe an algorithm that k-sorts an arbitrary array in $O(n \log k)$ time.
- (b) Prove that any comparison-based k-sorting algorithm requires $\Omega(n\log k)$ comparisons in the worst case.
- (c) Describe an algorithm that completely sorts an already k-sorted array in $O(n\log(n/k))$ time.
- (d) Prove that any comparison-based algorithm to completely sort a k-sorted array requires $\Omega(n\log(n/k))$ comparisons in the worst case.

Repeated Elements Problem

Convex Polygon Diameter (DH 6.8)

Show that the "Convex Polygon Diameter" algorithm is of linear-time complexity.

Convex Polygon Diameter (DH 6.8)

Show that the "Convex Polygon Diameter" algorithm is of linear-time complexity.

Linear-time of WHAT?

Time Complexity

An Simple Observation]

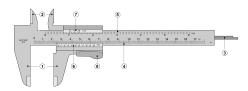
Definition (Line of Support)

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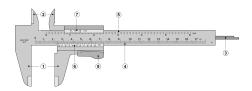
Definition (Antipodal)

Theorem

Rotating Caliper



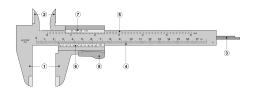
Rotating Caliper





"Computational Geometry" Ph.D Thesis, Michael Shamos, 1978

Rotating Caliper









"Solving Geometric Problems with the Rotating Calipers", 1983

Thank You!