

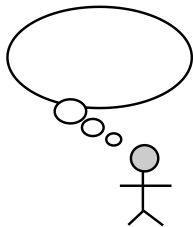
2-2 The Efficiency of Algorithms

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2018 年 04 月 02 日

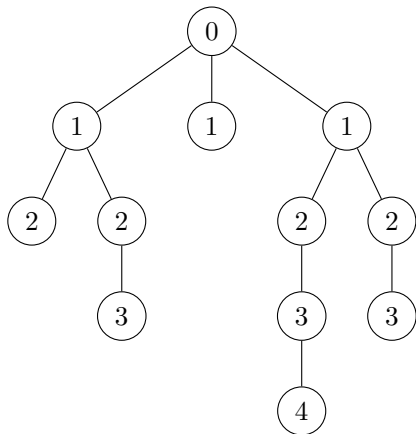
- (1) Diameter of Convex Polygon: $\Theta(n)$
- (2) Lower Bound for Sorting: $\Omega(n \lg n)$
- (3) Traversal over Trees: DFS/BFS ($\Theta(n)$)



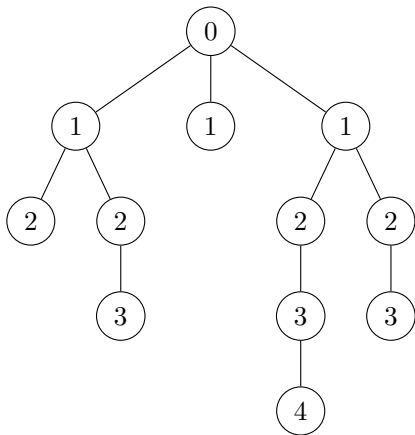
- (1) Diameter of Convex Polygon: $\Theta(n)$
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I have thought that ...

DH 4.2 (a): Sum of Depths

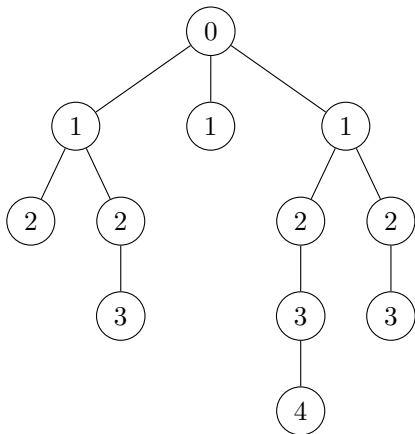


DH 4.2 (a): Sum of Depths



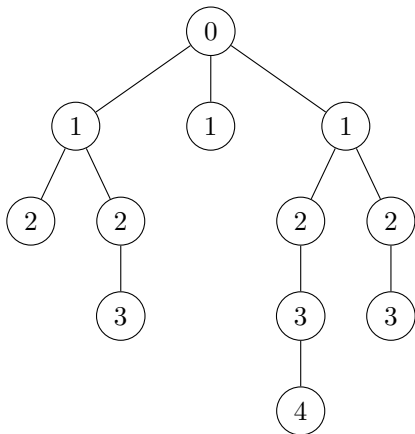
$$\text{sum-of-depths}(r) = \left\{ \sum_{v: \text{child of } r} \text{sum-of-depths}(v) + \text{depth of } r, \right.$$

DH 4.2 (a): Sum of Depths



$$\text{sum-of-depths}(r) = \begin{cases} \text{depth of } r, & r \text{ is a leaf} \\ \sum_{v: \text{child of } r} \text{sum-of-depths}(v) + \text{depth of } r, & \text{o.w.} \end{cases}$$

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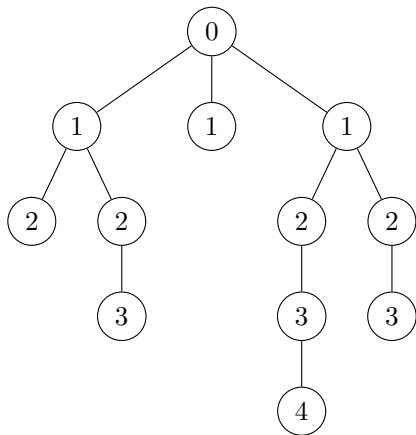


$$\text{sum-of-depths}(r, d) = \begin{cases} d, & r \text{ is a leaf} \\ \sum_{v: \text{child of } r} \text{sum-of-depths}(v, d+1) + d, & \text{o.w.} \end{cases}$$

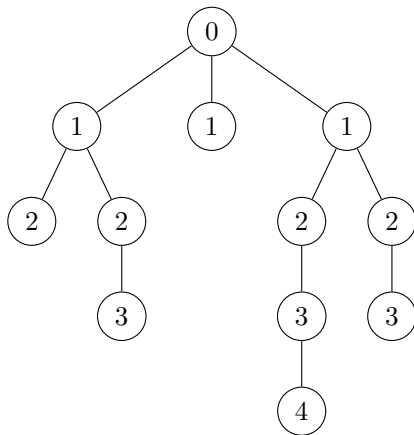
Algorithm 1 Calculate the sum of depths of all nodes of a tree T .

```
1: procedure SUM-OF-DEPTHS()  
2:   return SUM-OF-DEPTHS( $T, 0$ )  
  
3: procedure SUM-OF-DEPTHS( $r, depth$ )                                ▷  $r$ : 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

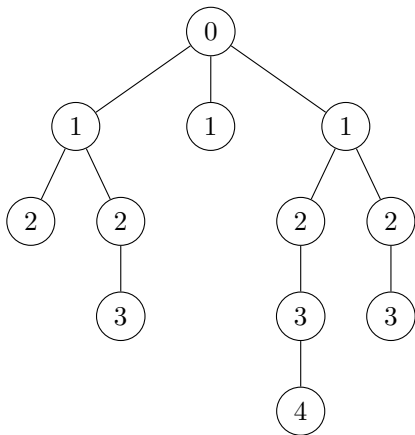


DH 4.2 (b): Number of Nodes at Depth K



$$\text{nodes-at-depth}(r, k) = \left\{ \begin{array}{l} \sum_{v: \text{child of } r} \text{nodes-at-depth}(v, k-1), \end{array} \right.$$

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

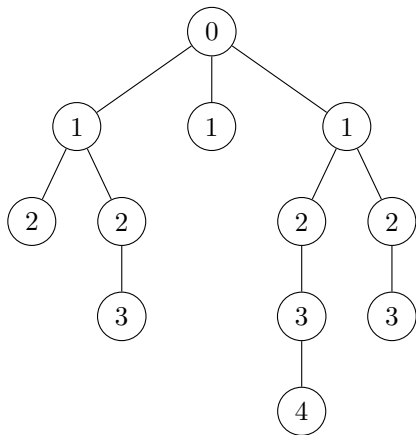


$$\text{nodes-at-depth}(r, k) = \begin{cases} 1, & k = 0 \\ 0, & k > 0 \wedge r \text{ is a leaf} \\ \sum_{v: \text{child of } r} \text{nodes-at-depth}(v, k-1), & \text{o.w.} \end{cases}$$

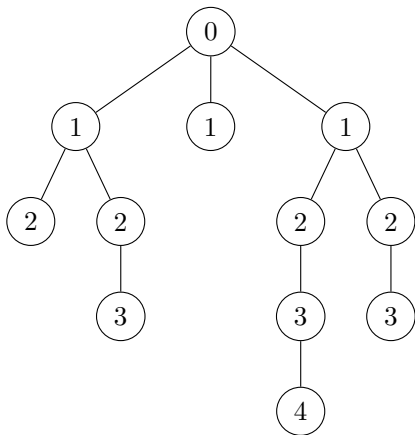
Algorithm 2 Count the number of nodes in T at depth K .

```
1: procedure NODES-AT-DEPTH()  
2:   return NODES-AT-DEPTH( $T, K$ )  
  
3: procedure NODES-AT-DEPTH( $r, k$ )                                ▷  $r$ : root of a tree  
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  $num$ 
```

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

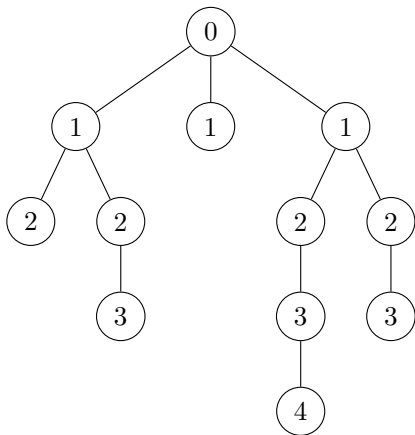


DH 4.2 (c): Any Leaf at an Even Depth?



$$\text{leaf-at-depth}(r, \textit{parity}) = \begin{cases} \sum_{v: \text{child of } r} (v, 1 - \textit{parity}), \end{cases}$$

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

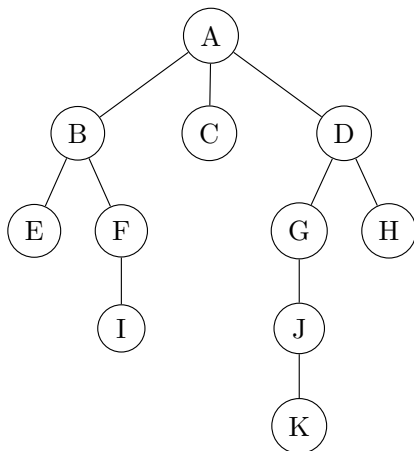


$$\text{leaf-at-depth}(r, \textit{parity}) = \begin{cases} 1 - \textit{parity}, & r \text{ is a leaf} \\ \sum_{v: \text{child of } r} (v, 1 - \textit{parity}), & \text{o.w.} \end{cases}$$

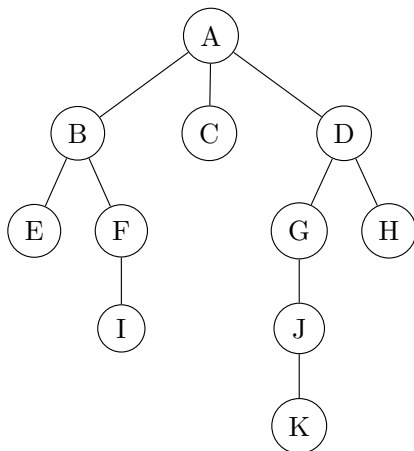
Algorithm 3 Check whether a tree T has any leaf at an even depth.

```
1: procedure LEAF-AT-EVEN-DEPTH()  
2:   return LEAF-AT-DEPTH( $T, even = 0$ )  
  
3: procedure LEAF-AT-DEPTH( $r, parity$ )                                ▷  $r$ : root of a tree  
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:      $result \leftarrow result \vee \text{LEAF-AT-DEPTH}(v, 1 - parity)$   
9:   return  $result$ 
```

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



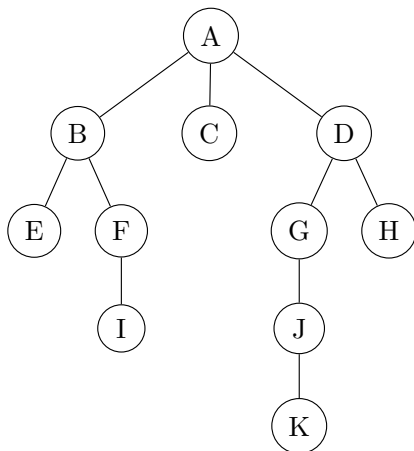
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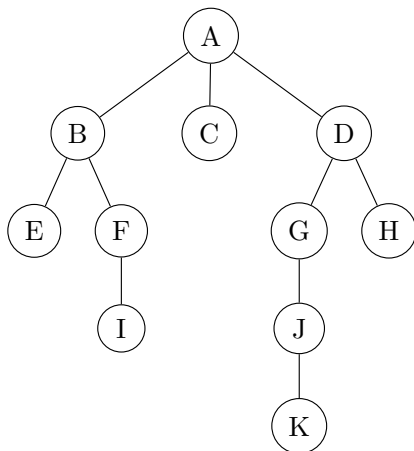
Algorithm 4 Calculate the sum of contents of nodes of a tree T at each depth.

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

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

Lower Bound for Comparison-based Sorting

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Lower Bound for Comparison-based Sorting (DH 6.13)

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

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Computational Model:

the only way to gain order info.

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

$$x/10$$

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Cost Model:

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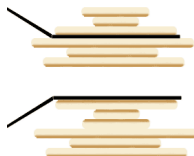
the critical operations to count

Computational Model:

the only way to gain order info.

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

$$x/10$$



“Bounds For Sorting By Prefix Reversal”, 1979

An argument from a student:

- ▶ Any comparison-based sorting algorithm can be considered to work by *putting elements into their final positions one by one*. (Take the last time one element is put into its final position.)

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- ▶ Any comparison-based sorting algorithm can be considered to work by *putting elements into their final positions one by one*. (Take the last time one element is put into its final position.)
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- ▶ Therefore, *the total number of comparisons* is at least

$$\sum_{i=1}^{n-1} \lg i = \Omega(n \lg n).$$



Is this lower bound proof for the comparison-based sorting problem correct?





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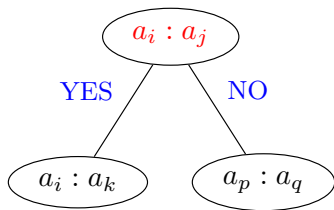


"This makes claims without justification."

— D.W.

Decision Tree Model

Decision Tree Model



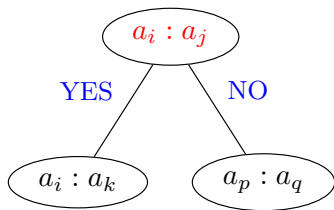
Nodes: comparisons $a_i : a_j$

$<, \leq, =, \geq, >$

Edges: two-way decisions (Y/N)

Leaves: possible permutations

Decision Tree Model



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

All the input elements are **distinct**.

$$a_i < a_j$$

Decision Tree Model

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

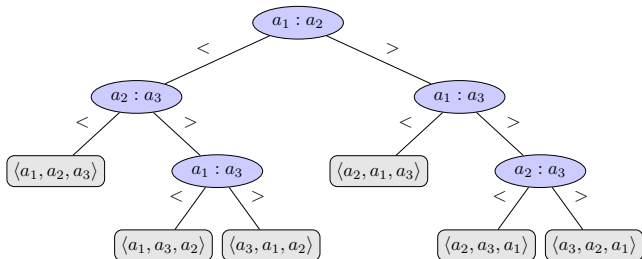
Decision Tree Model

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



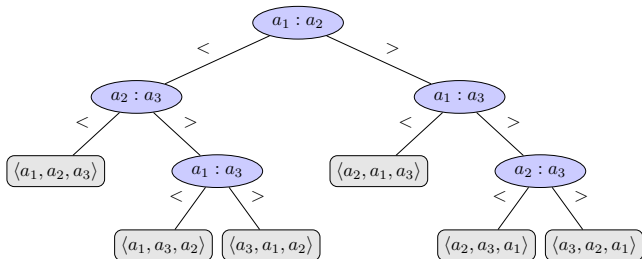
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Decision Tree Model

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



The decision tree for **insertion sort** on three elements.

Decision Tree Model

Any Comparison-based Sorting Algorithm modeled by A Decision Tree

Decision Tree Model

Any Comparison-based Sorting Algorithm modeled by 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( $A[j], A[i]$ )
```

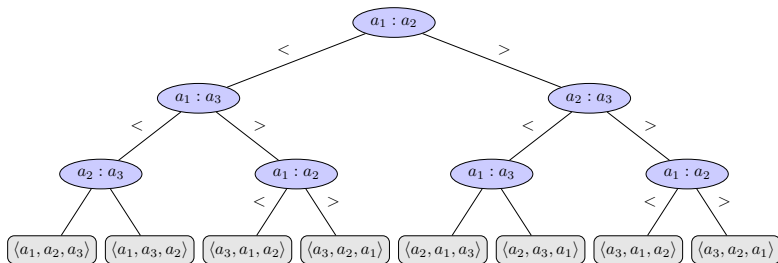
Decision Tree Model

Any Comparison-based Sorting Algorithm $\xrightarrow{\text{modeled by}}$ 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( $A[j], A[i]$ )
```

Decision Tree Model

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



The decision tree for **selection sort** on three elements.

Decision Tree Model

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

Decision Tree Model

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Decision Tree Model

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Worst-case Lower Bound of Comparison-based Sorting on inputs of size n
 $\xRightarrow{\text{modeled by}}$
The Minimum Height of All \mathcal{T} s

Decision Tree Model

Worst-case Lower Bound of Comparison-based Sorting on inputs of size n

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Decision Tree Model

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To be a full binary tree:

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

Decision Tree Model

Worst-case Lower Bound of Comparison-based Sorting on inputs of size n

modeled by 

The Minimum Height of All \mathcal{T} s

To be a full binary tree:

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

To be a correct sorting algorithm:

$$\# \text{ of leaves} \geq n!$$

Lower Bound for Comparison-based Sorting

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

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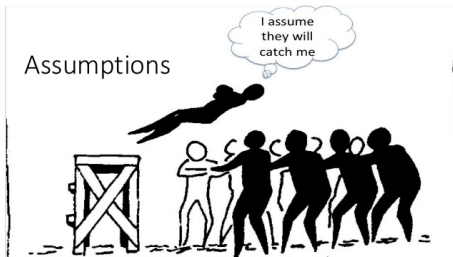
Stirling Formula (by *James Stirling*):

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

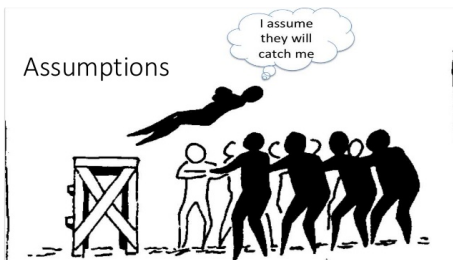


Looking Back

Looking Back



Looking Back



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

- (a) **Comparison-based** sorting algorithms.
- (b) All the input elements are **distinct**.

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.

Convex Polygon Diameter (DH 6.8)

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

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Linear-time of WHAT?

Time Complexity

Correctness

Correctness

[
An Simple Observation]

Correctness

Definition (Line of Support)

Correctness

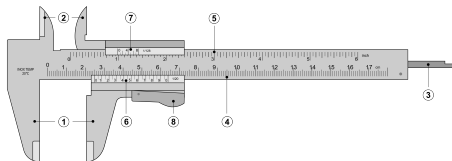
Definition (Line of Support)

Definition (Antipodal)

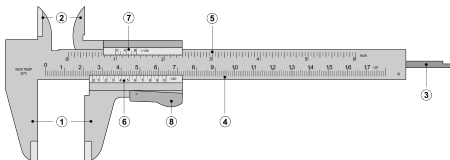
Correctness

Theorem

Rotating Caliper



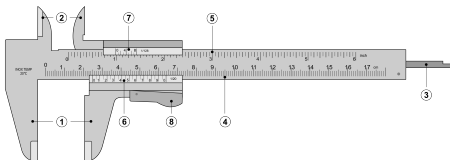
Rotating Caliper



“Computational Geometry”

Ph.D Thesis, Michael Shamos, 1978

Rotating Caliper



“Computational Geometry”
Ph.D Thesis, Michael Shamos, 1978



“Solving Geometric Problems with
the Rotating Calipers”, 1983

shortest distance





Repeated Elements Problem

Thank
You!