

## Lower Bound Theory

- **Lower bound:** an estimate of a number of operations needed to solve a given problem
- **Tight Lower Bound:**
  - There exists an algorithm with the same efficiency as the lower bound
- **Examples:**

Problem	Lower bound	Tightness
sorting (comparison-based)	$\Omega(n \log n)$	yes
searching in a sorted array	$\Omega(\log n)$	yes
n-digit integer multiplication	$\Omega(n)$	unknown
multiplication of n-by-n matrices	$\Omega(n^2)$	unknown

- **Methods of establishing lower bounds:**
  - Trivial lower bounds
    - Sorting
  - Information-theoretic arguments (decision trees)
    - Any comparison sorting algorithm (i.e., bubble sort)

- A convenient model of algorithms involving comparisons in which (Like sorting):
- Internal nodes represent comparisons
- Leaves represent outcomes
- Adversary arguments:
  - Merging two sorted lists
  - It's a game between the adversary and the (unknown) algorithm.
  - The adversary has the input and the algorithm asks questions to the adversary about the input.
  - The adversary tries to make the algorithm work the hardest by adjusting the input (consistently).
  - It wins the “game” after the lower bound time (lower bound proven) if it is able to come up with two different inputs.

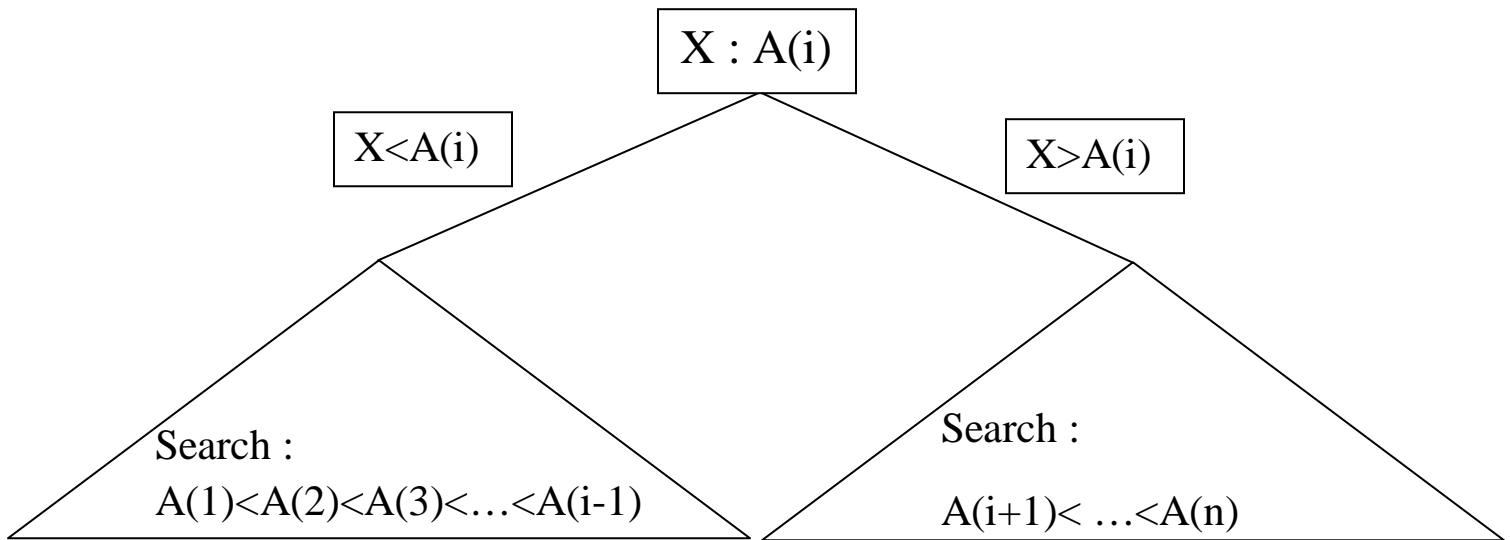
- **Searching in a sorted list:**

- Objective:  $A(1) < A(2) < A(3) < \dots < A(n)$

- Examples:

- Comparison-based search algorithms
    - Search list by comparing target element with list elements
    - Sequential search: order  $n$
    - Binary search: order  $\log_2 n$

- Comparison Tree:



- Let us denote  $TB(n)$  be the worst case for best algorithm.
    - $TB(n)$  = highest of the best tree = Shortest highest of trees
    - Each node will have  $n$  nodes

- Lemma: A tree of nodes and height  $h$ , then

$$h \geq \lceil \log (n + 1) \rceil - 1$$

- Then,  $TB(n) \geq \lceil \log (n + 1) \rceil - 1$

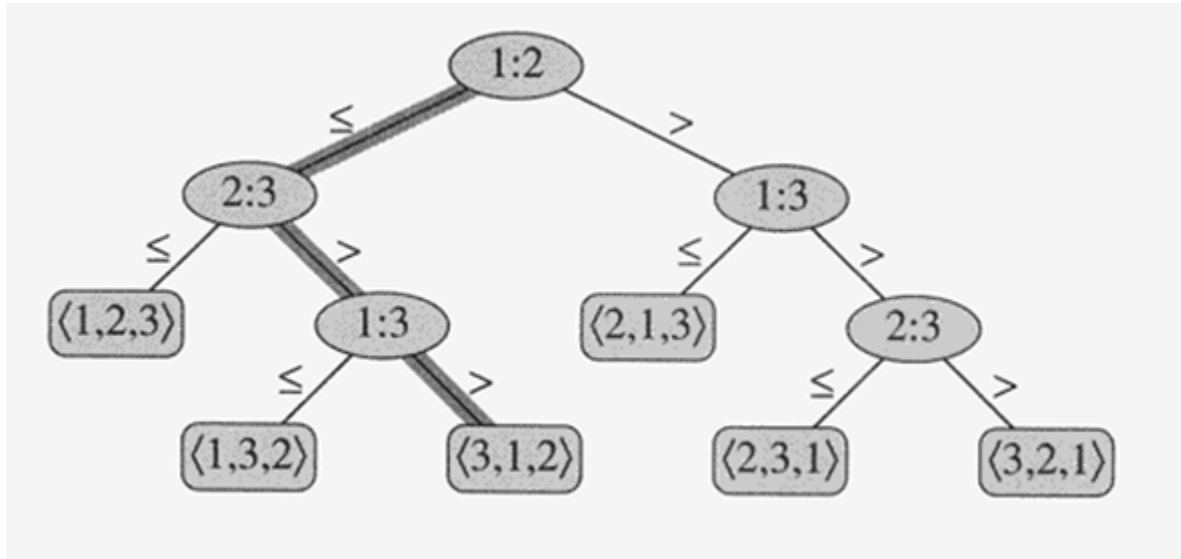
- **Finding the Minimum and Maximum**

- Let's consider the complexity of finding the largest and smallest elements. More formally, Given a sequence  $X = \langle x_1, x_2, \dots, x_n \rangle$  of  $n$  distinct numbers, find indices  $i$  and  $j$  such that  $x_i = \min(X)$  and  $x_j = \max(X)$ .
- How many comparisons do we need to solve this problem?
  - An upper bound of  **$2n - 3$**  is easy:
    - Find the minimum in  **$n - 1$  comparisons**, and then find the maximum of everything else in  **$n - 2$  comparisons**.
    - Similarly, a lower bound of  $n - 1$  is easy, since any algorithm that finds the min and the max certainly finds the max.
  - We **can improve** both the upper and the lower bound to:
$$\lceil 3n/2 \rceil - 2$$
  - The upper bound is established by the following algorithm:
    - Compare all  $\lceil n/2 \rceil$  consecutive pairs of elements  $x_{2i-1}$  and  $x_{2i}$
    - Put the smaller element into a set  $S$

- Put the larger element into a set L and if n is odd, put  $x_n$  into both L and S.
- Then find the smallest element of S and the largest element of L.
- The total number of comparisons is at most:
  - $\left\lceil \frac{n}{2} \right\rceil$ : Build S and L
  - $\left\lceil \frac{n}{2} \right\rceil - 1$ : Compute min S
  - $\left\lceil \frac{n}{2} \right\rceil - 1$ : Compute max L
  - $\left\lceil \frac{n}{2} \right\rceil + \left\lceil \frac{n}{2} \right\rceil - 1 + \left\lceil \frac{n}{2} \right\rceil - 1 = \lceil 3n/2 \rceil - 2$

- **Sorting**

- Decision tree for sorting 3 elements (Your textbook)



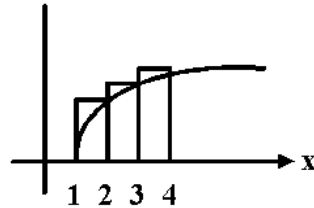
- To find the lower bound, we have to find the smallest depth of a binary tree.
- We have  $n!$  distinct permutations:  $n!$  leaf nodes in the binary decision tree.
- The balanced tree has the smallest depth which the lowest bound for sorting:

$$\lceil \log(n!) \rceil = \Omega(n \log n)$$

- **Method 1:**

$$\blacksquare \log(n!) = \log(n(n-1)\dots 1)$$

$$= \log 2 + \log 3 + \dots + \log n > \int_1^n \log x \, dx$$



$$= \log e \int_1^n \ln x \, dx$$

$$= \log e [x \ln x - x]_1^n$$

$$= \log e (n \ln n - n + 1)$$

$$= n \log n - n \log e + 1.44$$

$$\geq n \log n - 1.44n$$

$$= \Omega(n \log n)$$

$$\blacksquare \text{ lower bound for sorting: } \Omega(n \log n)$$

### ○ **Method 2:**

$$\blacksquare \text{ Using Sterling Approximation}$$

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

$$\log n! \approx \log \sqrt{2\pi} + \frac{1}{2} \log n + n \log \frac{n}{e}$$

$$\approx n \log n \approx \Omega(n \log n)$$

### ● **Merging two sorted lists:**



- Merge two sorted sequences A and B with lengths m and n.

- Binary decision tree:

There are  $\binom{m+n}{n}$  leaf nodes in the binary tree

- There are  $\binom{m+n}{n}$  ways !

- So the lower bound for merging:

$$\left\lceil \log \binom{m+n}{n} \right\rceil \leq m + n - 1 \text{ (conventional merging)}$$

- When m = n

$$\log \binom{m+n}{n} = \log \frac{(2n)!}{(n!)^2} = \log((2n)!) - 2\log n!$$

- Using Stirling approximation

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

$$\begin{aligned} \log((2n)!) &\approx 1/2 \log 4\pi + 1/2 \log n + 2n \log(2n/e) \\ &= 1/2 \log n + 2n \log 2 + 2n \log(n/e) + O(1) \end{aligned}$$

$$\log(n!) \approx 1/2 \log n + n \log(n/e) + O(1)$$

So,

$$\log \binom{m+n}{n} \approx 1/2 \log n + 2n \log 2 + 2n \log(n/e)$$

$$- \log n + 2n \log(n/e)$$

$$= 2n \log 2 - 1/2 \log n + O(1) = \Omega(n)$$

Thus,

$$\log \binom{m+n}{n} \approx \Omega(n)$$