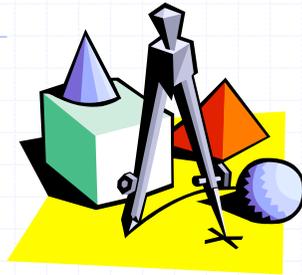


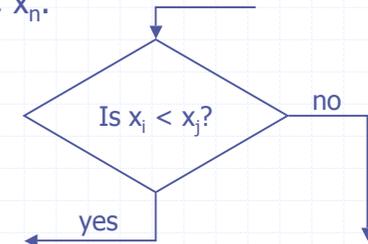
# Sorting Lower Bound



# Comparison-Based Sorting (§ 10.3)

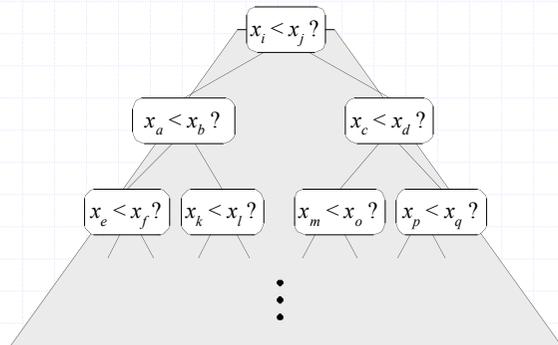


- ◆ Many sorting algorithms are comparison based.
  - They sort by making comparisons between pairs of objects
  - Examples: bubble-sort, selection-sort, insertion-sort, heap-sort, merge-sort, quick-sort, ...
- ◆ Let us therefore derive a lower bound on the running time of any algorithm that uses comparisons to sort  $n$  elements,  $x_1, x_2, \dots, x_n$ .



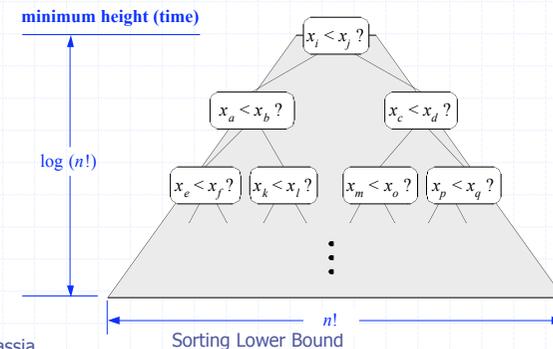
# Counting Comparisons

- ◆ Let us just count comparisons then.
- ◆ Each possible run of the algorithm corresponds to a root-to-leaf path in a **decision tree**



# Decision Tree Height

- ◆ The height of this decision tree is a lower bound on the running time
- ◆ Every possible input permutation must lead to a separate leaf output.
  - If not, some input ...4...5... would have same output ordering as ...5...4..., which would be wrong.
- ◆ Since there are  $n! = 1 * 2 * \dots * n$  leaves, the height is at least  $\log(n!)$



# The Lower Bound



- ◆ Any comparison-based sorting algorithms takes at least  $\log(n!)$  time
- ◆ Therefore, any such algorithm takes time at least

$$\begin{aligned} \log(n!) &= \log 1 + \log 2 + \dots + \log n \\ &\geq (n \log n) / 2 \end{aligned}$$

(by concavity, or integration)

- ◆ That is, any comparison-based sorting algorithm must require at least  $\Omega(n \log n)$  time.

# Summary of Sorting Algorithms

Algorithm	Time	Notes
selection-sort	$O(n^2)$	◆ in-place ◆ slow (good for small inputs)
insertion-sort	$O(n^2)$	◆ in-place ◆ slow (good for small inputs)
quick-sort	$O(n \log n)$ expected	◆ in-place, randomized ◆ fastest (good for large inputs)
heap-sort	$O(n \log n)$	◆ in-place ◆ fast (good for large inputs)
merge-sort	$O(n \log n)$	◆ sequential data access ◆ fast (good for huge inputs)