

# Computational Complexity

1. Time Complexity
  2. Space Complexity
- 

# Measuring the Performance of algorithm

Two important measures

1. Time Complexity
2. Space Complexity

# Running Time of Algorithm

- = How much time the algorithm uses **in terms of input size**
- = **# of certain operations** used in the algorithm where each operation take a constant time.  
**Like:** multiplications, additions, comparisons, assignments, shifts, ...etc
- ≠ # of seconds or minutes used by algorithm, because this depends on machine and technology (O.S., Prog. language, ...etc)

# Input Size vs. Input Type

- The running time is expressed in terms of input size and we concentrate our analysis on large inputs.
- Input types (arrays, lists, strings, integers, ..etc) is not an issue here.

# Algorithm Running Time

- Should be machine & technology independent.
- Should concentrate on the asymptotic times (large input size).
- Should concentrate on the main largest term (order of growth) and ignore the smaller ones.
- Sometimes we may even ignore the 1<sup>st</sup> constant (multiplicative) factor.
- The constant factors or the other smaller terms are important when comparing two algorithms of the same order of running time.
- Asymptotic Notations are used to describe asymptotic behavior of algorithm.

# Examples of Running Times (RT)

- Worst-case RT of Linear Search =  $\theta(n)$
- Worst-case RT of Binary Search =  $\theta(\log n)$ 
  - Average however is still =  $\theta(\log n)$
- RT of Selection Sort =  $n(n-1)/2 = \theta(n^2)$
- RT of Insertion Sort =  $\Omega(n)$  and  $O(n^2)$ 
  - Average however is still =  $\theta(n^2)$
- RT of Bottom Up Merge Sort =  $\theta(n \log n)$
- # of comparisons = RT

# Complexity of Running Time

- Could be
- Logarithmic =  $\theta(\log n)$
- Linear =  $\theta(n)$
- Quadratic =  $\theta(n^2)$
- Cubic =  $\theta(n^3)$
- Polynomial =  $\theta(n^k)$  ... all are efficient.

# How to compute the RT

- **Ugly:** by going through the code & counting iterations and operations
- **Beautiful:** by doing smart abstract analysis based on the idea of the algorithm



# Space Complexity

- Is defined to be the extra space used by the algorithm beside the space allocated to hold the input
- I.e., it is the work space used by the algorithm measured by the number of cells or words.

# Examples

- Linear Search uses  $\theta(1)$  (extra) space
- And so is Binary Search, Selection Sort, and Insertion Sort
- Merge Sort however uses  $\theta(n)$  extra space

# Optimal Algorithms

- These are algorithms whose worst-case performances meet the best-case performance of any algorithm that solves the same problem
- Optimal RT =  $\min \{ \text{RT of } A : A \text{ algorithm solves the problem} \}$

# Example

- Merge Sort is optimal among all comparisons-based sorting algorithms, because it uses  $\theta(n \log n)$ .
- **Theorem:** Any comparisons-based sorting algorithm uses  $\Omega(n \log n)$  cmps.
- There are however other non-cmp-based sorting algorithm that do better.

# Worst-case vs. Average-case

- Which one is better
  - low worst-case with high average or
  - low average with high worst-case?
- The average is the average!

# Input Size = $n$

- Sorting & searching:  $n = \#$  of elements in the array
- Graphs:  $n, m = \#$  of vertices & edges
- Integer Multiplications:  $n = \#$  of bits
- Cryptography:  $n = \#$  of bits