

# CSE5311 Design and Analysis of Algorithms

- This Class
  - What is an algorithm?
  - Asymptotic Analysis
  - Iterative algorithms
  - Recursive algorithms
- At the end of the class
  - Difference between an algorithm and a program
  - $O$ ,  $\Omega$ , and  $\Theta$  notations
    - How to use them
    - Determine complexity of a given algorithm
  - Write recurrence relations for your algorithms

Chapters 1 and 2

Algorithm Design *Kleinberg and Tardos*

# Course Syllabus

- **Review of Asymptotic Analysis and Growth of Functions**
- Trees, Heaps, and Graphs; and Recurrences.
- Greedy Algorithms:
  - Minimum spanning tree, Union-Find algorithms, Kruskal's Algorithm,
  - Clustering,
  - Huffman Codes, and
  - Multiphase greedy algorithms.
- Dynamic Programming:
  - Shortest paths, negative cycles, matrix chain multiplications, sequence alignment, RNA secondary structure, application examples.
- Network Flow:
  - Maximum flow problem, Ford-Fulkerson algorithm, augmenting paths, Bipartite matching problem, disjoint paths and application problems.
- NP and Computational tractability:
  - Polynomial time reductions; The Satisfiability problem; NP-Complete problems; and Extending limits of tractability.
- Approximation Algorithms, Local Search and Randomized Algorithms

# What are Algorithms ?

- An algorithm is a precise and unambiguous specification of a sequence of steps that can be carried out to solve a given problem or to achieve a given condition.
- An algorithm is a computational procedure to solve a well defined computational problem.
- An algorithm accepts some value or set of values as input and produces a value or set of values as output.
- An algorithm transforms the input to the output.
- Algorithms are closely intertwined with the nature of the data structure of the input and output values.

**Data structures are methods for representing the data models on a computer whereas data models are abstractions used to formulate problems.**

# What are these algorithms? Input? Output? Complexity?

**ALGO\_DO\_SOMETHING (A [1,...,n],1,n )**

- 1.**for** i  $\leftarrow$  1 to n-1
- 2.       small  $\leftarrow$  i;
- 3.       **for** j  $\leftarrow$  i+1 to n
- 4.             **if** A[j] < A[small] **then**
- 5.                 small  $\leftarrow$  j;
- 6.             temp  $\leftarrow$  A[small];
- 7.             A[small]  $\leftarrow$  A[i];
- 8.             A[i]  $\leftarrow$  temp;
- 9.**end**

**ALGO\_IMPROVED (A[1,...,n],i,n)**

- while** i < n
- **do** small  $\leftarrow$  i;
- **for** j  $\leftarrow$  i+1 to n
- **if** A[j] < A[small] **then**
- small  $\leftarrow$  j;
- temp  $\leftarrow$  A[small];
- A[small]  $\leftarrow$  A[i];
- A[i]  $\leftarrow$  temp;
- ALGO\_IMPROVED(A,i+1,n)
- End**

# Examples

- **Algorithms:**

An algorithm to sort a sequence of numbers into nondecreasing order.

**Application :** lexicographical ordering

An algorithm to find the shortest path from a source node to a destination node in a graph

**Application\_:** To find the shortest path from one city to another.

- **Data Models:**

**Lists, Trees, Sets, Relations, Graphs**

- **Data Structures :**

**Linked List** is a data structure used to represent a List

**Graph** is a data structure used to represent various cities in a map.

# SELECTION SORT Algorithm (*Iterative method*)

Procedure **SELECTION\_SORT** (A [1,...,n])

Input : unsorted array A

Output : Sorted array A

```
1.  for i ← 1 to n-1
2.      small ← i;
3.      for j ← i+1 to n
4.          if A[j] < A[small] then
5.              small ← j;
6.          temp ← A[small];
7.          A[small] ← A[i];
8.          A[i] ← temp;
9.  end
```

Example: Given sequence

5 2 4 6 1 3

i=1 1 2 4 6 5 3

i=2 1 2 4 6 5 3

i=3 1 2 3 6 5 4

i=4 1 2 3 4 5 6

```
1.   for i ← 1 to n-1
2.       small ← i;
3.           for j ← i+1 to n
4.               if A[j] < A[small] then
5.                   small ← j;
6.           temp ← A[small];
7.           A[small] ← A[i];
8.           A[i] ← temp;
9.   end
```

## Complexity:

The statements 2,6,7,8, and 5 take  $O(1)$  or constant time.  
The outerloop 1-9 is executed  $n-1$  times and the inner loop 3-5 is executed  $(n-i)$  times.

The upper bound on the time taken by all iterations as  $i$  ranges from 1 to  $n-1$  is given by,  **$O(n^2)$**

- Study of algorithms involves,
  - **designing algorithms**
  - **expressing algorithms**
  - **algorithm validation**
  - **algorithm analysis**
  - **Study of algorithmic techniques**

# ***Algorithms and Design of Programs***

- ***An algorithm is composed of a finite set of steps,***
    - \* **each step may require one or more operations,**
    - \* **each operation must be definite and effective**
  - ***An algorithm,***
    - \* **is an abstraction of an actual program**
    - \* **is a computational procedure that terminates**
- \***A program is an expression of an algorithm in a programming language.**
- \***Choice of proper data models and hence data structures is important for expressing algorithms and implementation.**

- We evaluate the performance of algorithms based on **time** (CPU-time) and **space** (semiconductor memory) required to implement these algorithms. However, both these are expensive and a computer scientist should endeavor to minimize time taken and space required.

- The time taken to execute an algorithm is dependent on one or more of the following,
  - **number of data elements**
  - **the degree of a polynomial**
  - **the size of a file to be sorted**
  - **the number of nodes in a graph**

# Asymptotic Notations

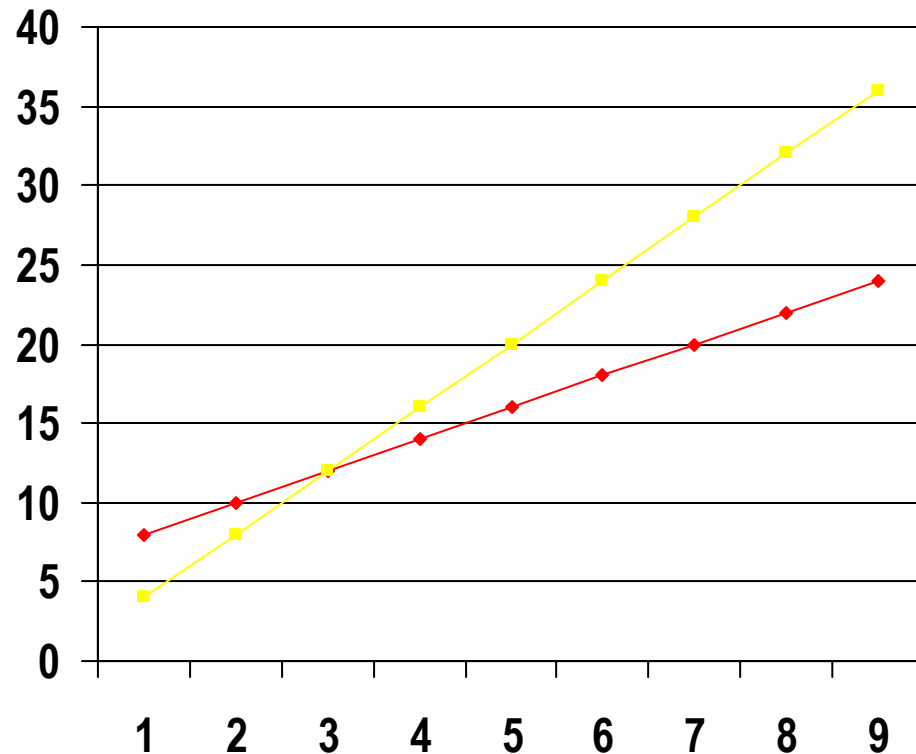
## – O-notation

### » Asymptotic upper bound

- A given function  $f(n)$ , is  $O(g(n))$  if there exist positive constants  $c$  and  $n_0$  such that  $0 \leq f(n) \leq c g(n)$  for all  $n \geq n_0$ .
- $O(g(n))$  represents a set of functions, and  $O(g(n)) = \{f(n): \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq c g(n) \text{ for all } n \geq n_0\}$ .

# O Notation

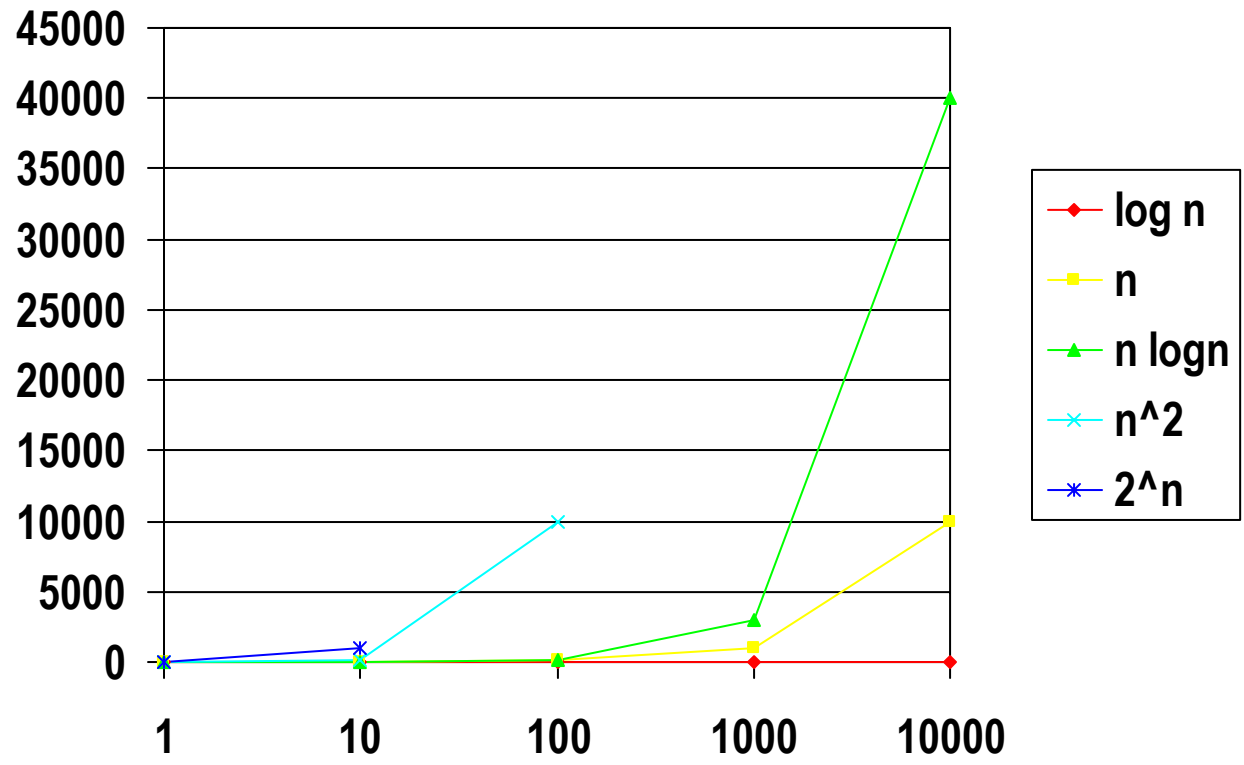
$f(n)$ , is  $O(g(n))$  if there exist positive constants  $c$  and  $n_0$  such that  $0 \leq f(n) \leq c g(n)$  for all  $n \geq n_0$ .



—◆—  $f(n) = 2n+6$   
—■—  $cg(n) = 4n$

$$c = 4$$

$$n_0 = 3.5$$



## $\Omega$ -notation

### *Asymptotic lower bound*

- A given function  $f(n)$ , is  $\Omega(g(n))$  if there exist positive constants  $c$  and  $n_0$  such that

$$0 \leq c g(n) \leq f(n) \text{ for all } n \geq n_0.$$

- $\Omega(g(n))$  represents a set of functions, and

$\Omega(g(n)) = \{f(n): \text{there exist positive constants}$

$c \text{ and } n_0 \text{ such that } 0 \leq c g(n) \leq f(n) \text{ for all } n \geq n_0\}$

## $\Theta$ -notation

### *Asymptotic tight bound*

- A given function  $f(n)$ , is  $\Theta(g(n))$  if there exist positive constants  $c_1$ ,  $c_2$ , and  $n_0$  such that

$$0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0.$$

- $\Theta(g(n))$  represents a set of functions, and

$$\Theta(g(n)) = \{f(n) : \text{there exist positive constants } c_1, c_2, \text{ and } n_0 \text{ such that } 0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0.\}$$

*$O$ ,  $\Omega$ , and  $\Theta$  correspond (loosely) to “ $\leq$ ”, “ $\geq$ ”, and “ $=$ ”.*

# Presenting algorithms

- **Description** : The algorithm will be described in English, with the help of one or more examples
- **Specification** : The algorithm will be presented as pseudocode  
(We don't use any programming language)
- **Validation** : The algorithm will be proved to be correct for all problem cases
- **Analysis**: The running time or time complexity of the algorithm will be evaluated

# SELECTION SORT Algorithm (*Iterative method*)

**Procedure SELECTION\_SORT (A [1,...,n])**

**Input : unsorted array A**

**Output : Sorted array A**

```
1.   for i ← 1 to n-1
2.       small ← i;
3.       for j ← i+1 to n
4.           if A[j] < A[small] then
5.               small ← j;
6.       temp ← A[small];
7.       A[small] ← A[i];
8.       A[i] ← temp;
9.   end
```

# Recursive Selection Sort Algorithm

Given an array  $A[i, \dots, n]$ , selection sort picks the smallest element in the array and swaps it with  $A[i]$ , then sorts the remainder  $A[i+1, \dots, n]$  recursively.

Example :

Given A [26, 93, 36, 76, 85, **09**, 42, 64]

Swap 09 with 23 --  $A[1] = 09$ ;  $A[2, \dots, 8] = [93, 36, 76, 85, 26, 42, 64]$

Swap 26 with 93 --  $A[1, 2] = [09, 26]$ ;  $A[3, \dots, 8] = [36, 76, 85, 93, 42, 64]$

No swapping --  $A[1, 2, 3] = [09, 26, 36]$ ;  $A[4, \dots, 8] = [76, 85, 93, 42, 64]$

Swap 42 with 76 --  $A[1, \dots, 4] = [09, 26, 36, 42]$ ;  $A[5, \dots, 8] = [85, 93, 76, 64]$

Swap 64 with 85 --  $A[1, \dots, 5] = [09, 26, 36, 42, 64]$ ;  $A[6, 7, 8] = [93, 76, 85]$

Swap 76 with 93 --  $A[1, \dots, 6] = [09, 26, 36, 42, 64, 76]$ ;  $A[7, 8] = [93, 85]$

Swap 85 with 93 --  $A[1, \dots, 7] = [09, 26, 36, 42, 64, 76, 85]$ ;  $A[8] = 93$

Sorted list :  $A[1, \dots, 8] = [09, 26, 36, 42, 64, 76, 85, 93]$

Procedure **RECURSIVE\_SELECTION\_SORT** (A[1,...,n],i,n)

**Input : Unsorted array A**

**Output : Sorted array A**

```
while i < n
  do small ← i;
    for j ← i+1 to n
      if A[j] < A[small] then
        small ← j;
    temp ← A[small];
    A[small] ← A[i];
    A[i] ← temp;
    RECURSIVE_SELECTION_SORT(A,i+1,n)

End
```

## Analysis of Recursive selection sort algorithm

**Basis:** If  $i = n$ , then only the last element of the array needs to be sorted, takes  $\Theta(1)$  time.

Therefore,  $T(1) = a$ , a constant

**Induction :** if  $i < n$ , then,

1. we find the smallest element in  $A[i, \dots, n]$ ,  
takes at most  $(n-1)$  steps

    swap the smallest element with  $A[i]$ , one step  
    recursively sort  $A[i+1, \dots, n]$ , takes  $T(n-1)$  time

Therefore,  $T(n)$  is given by,

$$T(n) = T(n-1) + b \cdot n \quad (1)$$

It is required to solve the recursive equation,

$$T(1) = a; \text{ for } n = 1$$

$$T(n) = T(n-1) + b n; \text{ for } n > 1, \text{ where } b \text{ is a constant}$$

$$T(n-1) = T(n-2) + (n-1)b \quad (2)$$

$$T(n-2) = T(n-3) + (n-2)b \quad (3)$$

...

$$T(n-i) = T(n-(i+1)) + (n-i)b \quad (4)$$

Using (2) in (1)

$$\begin{aligned} T(n) &= T(n-2) + b [n+(n-1)] \\ &= T(n-3) + b [n+(n-1)+(n-2)] \\ &= T(n-(n-1)) + b [n+(n-1)+(n-2) + \dots + (n-(n-2))] \end{aligned}$$

$$T(n) = O(n^2)$$

## Questions:

- What is an algorithm?
- Why should we study algorithms?
- Why should we evaluate running time of algorithms?
- What is a recursive function?
- What are the basic differences among  $O$ ,  $\Omega$ , and  $\Theta$  notations?
- Did you understand selection sort algorithm and its running time evaluation?
- Can you write pseudocode for selecting the largest element in a given array?  
Please write the algorithm in the class.

Home work: Please read

Chapters 1 and 2, *Algorithm Design Kleinberg and Tardos*