

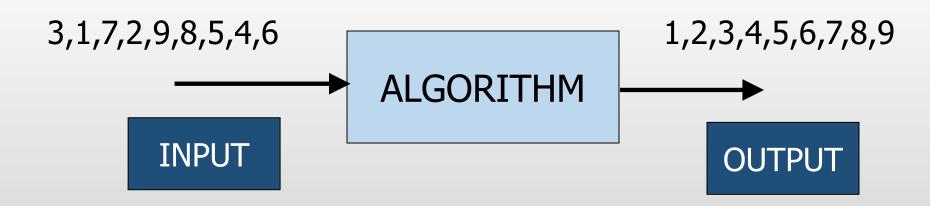
# IT2070 – Data Structures and Algorithms

Lecture 05
Introduction to Algorithms



#### **ALGORITHMS**

 Algorithm is any well defined computational procedure that takes some value or set of values as input and produce some value or set of values as output.





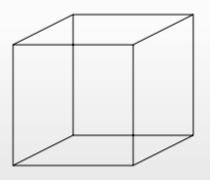
# ALGORITHM (Contd.)

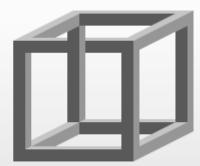
- 1.Get the smallest value from the input.
- 2. Remove it and output.
- 3. Repeat above 1,2 for remaining input until there is no item in the input.



# Properties of an Algorithm.

- Be correct.
- Be unambiguous.
- Give the correct solution for all cases.
- Be simple.
- It must terminate.





Necker\_cube\_and\_impossible\_cube

Source:http://en.wikipedia.org/wiki/Ambiguity#Mathematical\_i
nterpretation\_of\_ambiguity



#### Applications of Algorithms

- Data retrieval
- Network routing
- Sorting
- Searching
- Shortest paths in a graph



#### Pseudocode

- Method of writing down a algorithm.
- Easy to read and understand.
- Just like other programming language.

- More expressive method.
- Does not concern with the technique of software engineering.



## Pseudocode Conventions.

- English.
- Indentation.
- Separate line for each instruction.
- Looping constructs and conditional constructs.
- // indicate a comment line.
- = indicate the assignment.



#### Pseudocode Conventions.

- Array elements are accessed by specifying the array name followed by the index in the square bracket.
- The notation ".." is used to indicate a range of values within the array.

Ex:

A[1..i] indicates the sub array of A consisting of elements A[1], A[2], ..., A[i].



# Analysis of Algorithms

Idea is to predict the resource usage.

- Memory
- Logic Gates
- Computational Time

Why do we need an analysis?

- To compare
- Predict the growth of run time



# Worst, Best and Average case.

Running time will depend on the chosen instance characteristics.

#### Best case:

Minimum number of steps taken on any instance of size n.

#### Worst case:

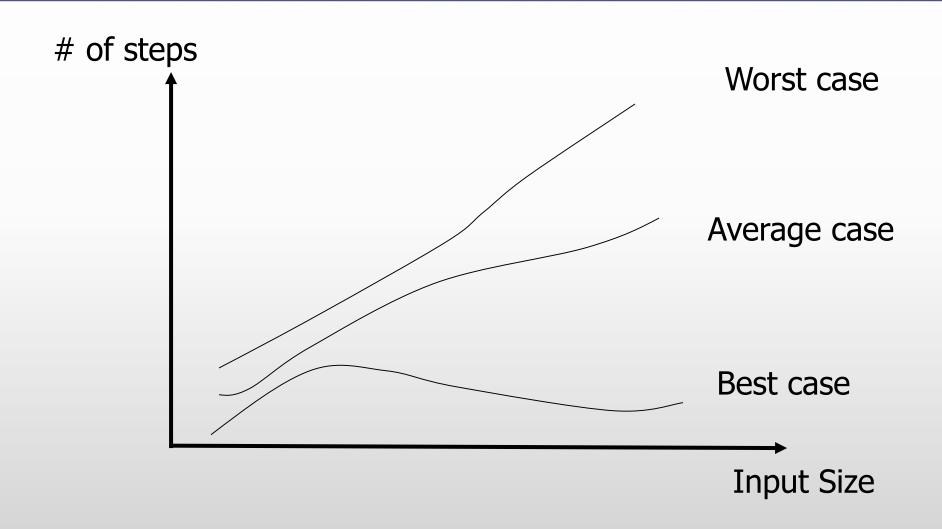
Maximum number of steps taken on any instance of size n.

#### Average case:

An average number of steps taken on any instance of size n.



# Worst, Best and Average case (Contd.)





# Analysis Methods

- Operation Count Methods
- Step Count Method(RAM Model)
- Exact Analysis
- Asymptotic Notations



## Operation count

- Methods for time complexity analysis.
- Select one or more operations such as add, multiply and compare.
- Operation count considers the **time spent on chosen operations** but not all.



# Step Count (RAM Model)

- Assume a generic one processor.
- Instructions are executed one after another, with no concurrent operations.
- +, -, =, it takes exactly one step.
- Each memory access takes exactly 1 step.
- Running Time = Sum of the steps.



## RAM Model Analysis.

# Example1: n = 100 1step n = n + 100 2steps Print n 1step Steps = 4

#### Example2:

$$sum = 0$$

for i = 1 to n

to 
$$n$$
  $n+1$  comparisons  $n$  additions  $n$  additions  $n$  additions

1 assignment

*n*+1 assignments

Steps = 
$$6n+3$$



• Using RAM model analysis, find out the no of steps needed to display the numbers from 1 to 10.

i = 1 → 1 step  
While i <=10 → 11 steps  
print i → 10 steps  

$$i = i + 1$$
 → 10 + 10 = 20 steps

$$Steps = 42$$



• Using RAM model analysis, find out the no of steps needed to display the numbers from 10 to 20.

```
i = 10 → 1 step

While i <= 20 → 12 steps (Hint:20 – 10 + 2 = 12)

print i → 11 steps

i = i + 1 → 11 + 11 = 22 steps
```

$$Steps = 46$$



• Using RAM model analysis, find out the no of steps needed to display the even numbers from 10 to 20.

for i = 10 to 20 
$$\rightarrow$$
 (12+ 12 + 11) steps = 35 steps  
if i % 2 == 0  $\rightarrow$  2 \* 11 = 22 steps  
print i  $\rightarrow$  6 steps

$$Steps = 63$$



#### Problems with RAM Model

Differ number of steps with different architecture.

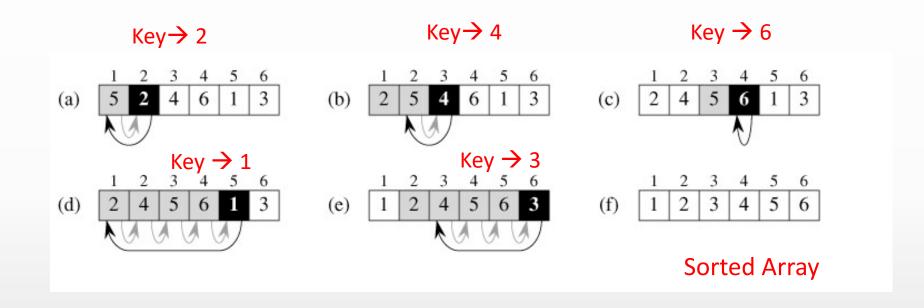
eg: sum = sum + A[i] is a one step in the CISC processor.

It is difficult to count the exact number of steps in the algorithm.

eg: See the insertion sort, efficient algorithm for sorting small number of elements.



#### Insertion sort



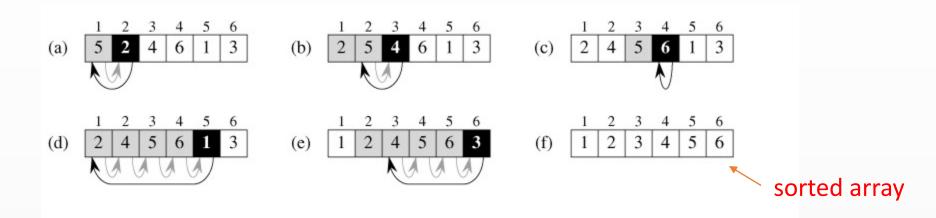


## Pseudocode for insertion sort.

#### **INSERTION-SORT(A) 1 for** j **=** 2 **to A.**length key = A[j]// Insert A[j] into the sorted sequence A[1..j-1] i = j - 1While i > 0 and A[i] > key 5 6 A[i+1] = A[i]i = i-18 A[i+1] = key



## Insertion sort - Example



- (a)-(e) The iterations of the for loop  $\rightarrow$  lines 1-8.
- In each iteration, the black rectangle holds the key taken from A[j],
- Key is compared with the values in shaded rectangles to its left → line 5.
- Shaded arrows show array values moved one position to the right  $\rightarrow$  line 6,
- Black arrows indicate where the key is moved to → line 8.



# Exact analysis of Insertion sort

 Time taken for the algorithm will depend on the input size (number of elements of the array)

#### **Running Time (Time complexity):**

This is the number of primitive operations or steps executed through an algorithm given a particular input.



# Running Time: T(n)

|   | INSERTION-SORT(A)                                  | Cost                  | Times                           |  |  |
|---|--|-----------------------|---------------------------------|--|--|
| 1 | for j = 2 to A.length                              | c <sub>1</sub>        | n                               |  |  |
| 2 | key = A[j]   | c <sub>2</sub>        | n-1                             |  |  |
| 3 | // Insert A[j] into the sorted // sequence A[1j-1] | 0                     | n-1                             |  |  |
| 4 | i = j — 1  | C <sub>4</sub>        | n-1                             |  |  |
| 5 | While i > 0 and A[i] > key                         | <b>c</b> <sub>5</sub> | $\sum_{j=2}^{n} \mathbf{t}_{j}$ |  |  |
| 6 | A[i+1] = A[i]                                      | c <sub>6</sub>        | $\sum_{j=2}^{n} (t_{j} - 1)$    |  |  |
| 7 | i = i-1  | c <sub>7</sub>        | $\sum_{j=2}^{n} (t_{j} - 1)$    |  |  |
| 8 | A[i+1] = key                                       | c <sub>8</sub>        | n-1                             |  |  |

 $i^{th}$  line takes time  $c_i$  where  $c_i$  is a constant.

For each j=2,3,...,n,  $t_j$  be the number of times the while loop is executed for that value of j



# Running Time(contd.)

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1) + c_5 \sum_{j=2}^{n} t_j$$

$$+ c_6 \sum_{j=2}^{n} (t_j - 1) + c_7 \sum_{j=2}^{n} (t_j - 1) + c_8 (n-1)$$

- Best Case (Array is in sorted order)
  - $T(n) \rightarrow an+b$
- Worst Case (Array is in reverse sorted order)
  - $T(n) \rightarrow cn^2 + dn + e$



## Worst Case $T(n) \rightarrow cn^2 + dn + e$

*Worst case:* The array is in reverse sorted order.

- Always find that A[i] > key in while loop test.
- Have to compare key with all elements to the left of the jth position ⇒ compare with j − 1 elements.
- Since the while loop exits because i reaches 0, there's one additional test after the j-1 tests  $\Rightarrow t_j=j$ .

• 
$$\sum_{j=2}^{n} t_j = \sum_{j=2}^{n} j$$
 and  $\sum_{j=2}^{n} (t_j - 1) = \sum_{j=2}^{n} (j - 1)$ .

•  $\sum_{j=1}^{n} j$  is known as an *arithmetic series*, and equation (A.1) shows that it equals  $\frac{n(n+1)}{2}$ .



## Worst Case $T(n) \rightarrow cn^2 + dn + e$

- Since  $\sum_{j=2}^{n} j = \left(\sum_{j=1}^{n} j\right) 1$ , it equals  $\frac{n(n+1)}{2} 1$ . [The parentheses around the summation are not strictly necessary. They are there for clarity, but it might be a good idea to remind the students that the meaning of the expression would be the same even without the parentheses.]
- Letting k = j 1, we see that  $\sum_{j=2}^{n} (j 1) = \sum_{k=1}^{n-1} k = \frac{n(n-1)}{2}$ .
- · Running time is

$$T(n) = c_1 n + c_2 (n-1) + c_4 (n-1) + c_5 \left(\frac{n(n+1)}{2} - 1\right)$$

$$+ c_6 \left(\frac{n(n-1)}{2}\right) + c_7 \left(\frac{n(n-1)}{2}\right) + c_8 (n-1)$$

$$= \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2}\right) n^2 + \left(c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8\right) n$$

$$- \left(c_2 + c_4 + c_5 + c_8\right).$$

• Can express T(n) as  $an^2 + bn + c$  for constants a, b, c (that again depend on statement costs)  $\Rightarrow T(n)$  is a quadratic function of n.



## Asymptotic Notations

- RAM Model has some problems.
- Exact analysis is very complicated.

#### Therefore we move to asymptotic notation

- Here we focus on determining the biggest term in the complexity function.
- Sufficiently large size of n.



# Asymptotic Notations(Contd.)

• There are three notations.

- **O** Notation
- **⊕** Notation
- $\Omega$  Notation



# Big O - Notation

- Introduced by Paul Bechman in 1892.
- We use Big O-notation to give an upper bound on a function.

#### **Definition:**

 $O(g(n)) = \{ f(n) : there exist positive constants c and n<sub>o</sub> such that$ 

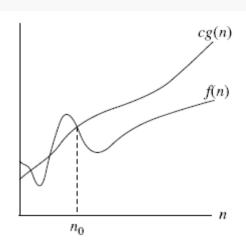
$$0 \le f(n) \le cg(n)$$
 for all  $n \ge n_o$ .

Eg: What is the big O value of f(n)=2n + 6?

$$c = 4$$
  
 $n_0 = 3$ 

g(n)=n therefore

$$f(n) = O(n)$$



g(n) is an *asymptotic upper bound* for f(n).

If  $f(n) \in O(g(n))$ , we write f(n) = O(g(n))



# Back to the example

#### • Alternative calculation:

|                    | cost                  | times |
|--------------------|-----------------------|-------|
| sum = 0            | $c_1$                 | 1     |
| for $i = 1$ to $n$ | $c_2$                 | n+1   |
| sum = sum + A[i]   | <i>c</i> <sub>3</sub> | n     |

$$T(n) = c_1 + c_2 (n+1) + c_3 n$$

$$= (c_1 + c_2) + (c_2 + c_3) n$$

$$= c_4 + c_5 n \rightarrow O(n)$$

Proof:  $c_4 + c_5 n \le c n \rightarrow \text{TRUE for } n \ge 1 \text{ and } c \ge c_4 + c_5$ 



# Big O - Notation(Contd.)

Assignment (s = 1)

Addition (s+1)

Multiplication (s\*2)

Comparison (S<10)

O(1)



• Find the Big O value for following fragment of code.

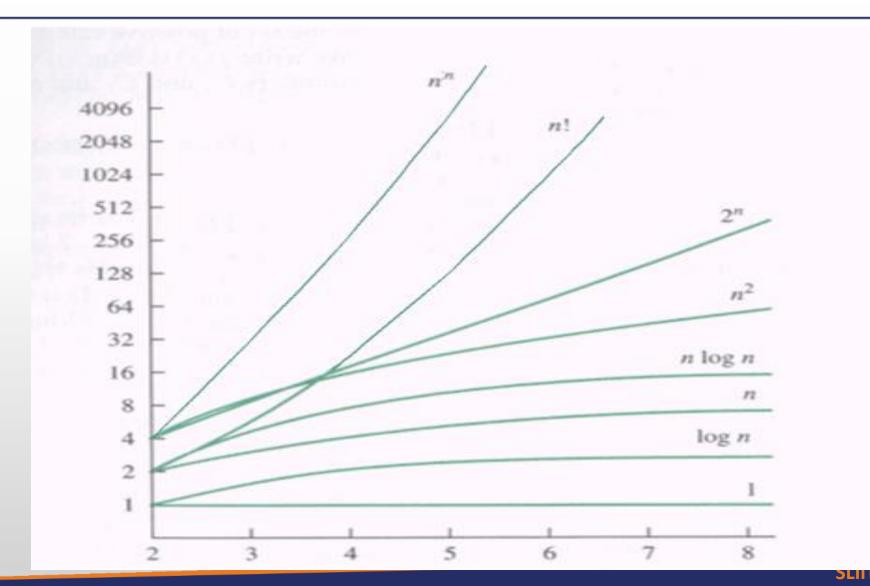
```
for i = 1 to n

for j = 1 to i

Print j
O(n^2)
```



## Graphs of functions





| n    | logn | n     | nlogn  | $n^2$     | $n^3$         | $2^n$                   |
|------|------|-------|--------|-----------|---------------|-------------------------|
| 4    | 2    | 4     | 8      | 16        | 64            | 16                      |
| 8    | 3    | 8     | 24     | 64        | 512           | 256                     |
| 16   | 4    | 16    | 64     | 256       | 4,096         | 65,536                  |
| 32   | 5    | 32    | 160    | 1,024     | 32,768        | 4,294,967,296           |
| 64   | 6    | 64    | 384    | 4,094     | 262,144       | 1.84 * 1019             |
| 128  | 7    | 128   | 896    | 16,384    | 2,097,152     | $3.40*10^{38}$          |
| 256  | 8    | 256   | 2,048  | 65,536    | 16,777,216    | 1.15 * 10 <sup>77</sup> |
| 512  | 9    | 512   | 4,608  | 262,144   | 134,217,728   | 1.34 * 10154            |
| 1024 | 10   | 1,024 | 10,240 | 1,048,576 | 1,073,741,824 | $1.79 * 10^{308}$       |



# Big O – Notation(Contd.)

- Find the Big O value for the following functions.
  - (i)  $T(n) = 3 + 5n + 3n^2$
  - (ii)  $f(n)= 2^n + n^2 + 8n + 7$
  - (iii) T(n) = n + logn + 6

#### **Answers:**

- (i)  $O(n^2)$
- (ii)  $O(2^n)$
- (iii) O(n)



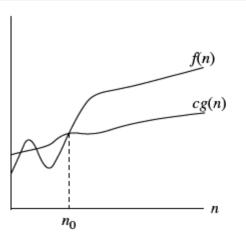
#### $\Omega$ - Notation

Provides the lower bound of the function.

#### **Definition:**

 $\Omega(g(n)) = \{ f(n) : \text{there exist positive constants c and } n_0 \text{ such that } 0 : \le f(n) \text{ for all } n \ge n_0 \}$ 

$$0 \le cg(n)$$



g(n) is an *asymptotic lower bound* for f(n).

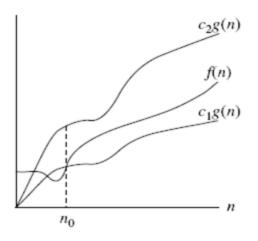


#### Θ - Notation

• This is used when the function f can be bounded both from above and below by the same function g.

#### **Definition:**

 $\Theta(g(n)) = \{ f(n): \text{ there exist positive constant } c_1, c_2, \text{ and } n_0 \text{ such that } 0 \le c_1 g(n) \le f(n) \le c_2 g(n) \text{ for all } n \ge n_0 \}$ 



g(n) is an *asymptotically tight bound* for f(n).



# Summary

- What is an algorithm?
- Properties of an algorithm.
- Design methods.
- Pseudocode.
- Analysis(Operation count & Step count, RAM model).
- Insertion Sort.
- Asymptotic Notation



#### References

• T.H. Cormen, C.E. Leiserson, R.L. Rivest, Clifford Stein Introduction to Algorithms, 3<sup>rd</sup> Edition, MIT Press, 2009.