

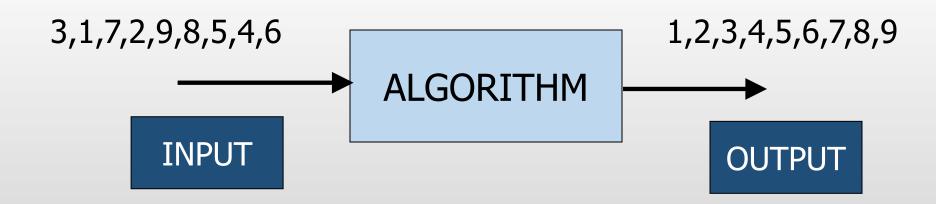
# IT2070 – Data Structures and Algorithms

Lecture 06
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.





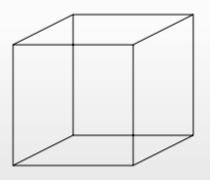
## ALGORITHM (Contd.)

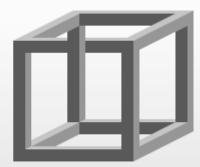
- 1.Get the smallest value from the input.
- 2. Remove it and output.
- 3. Repeat above 1,2 for remaining input until there is no item in the input.



## 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:

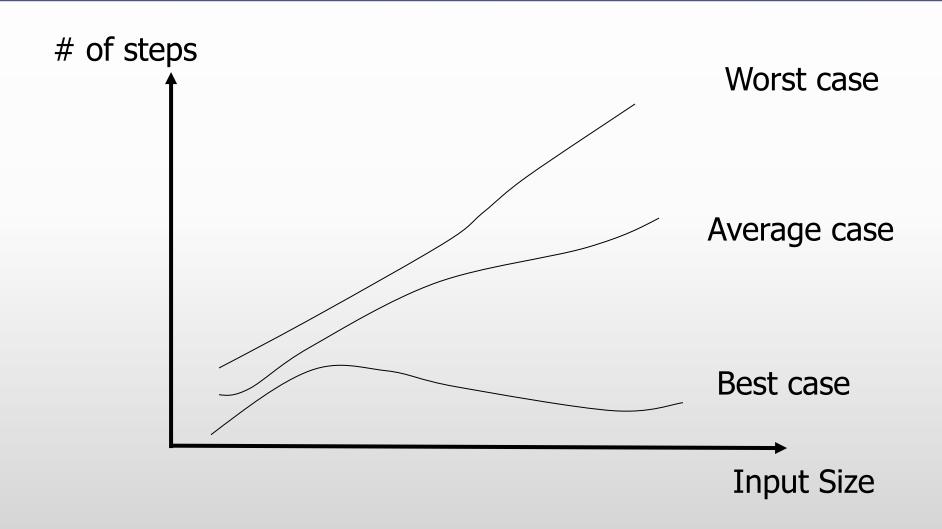
Maximum number of steps taken on any instance of size n.

#### 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 = 1001step n = n + 1002steps

1step

Steps = 4

Print n

#### Example2:

$$sum = 0$$

for i = 1 to n

*n*+1 comparisons *n* additions

1 assignment

*n*+1 assignments

n memory accesses

$$sum = sum + A[i]$$
 n assignments n additions

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.

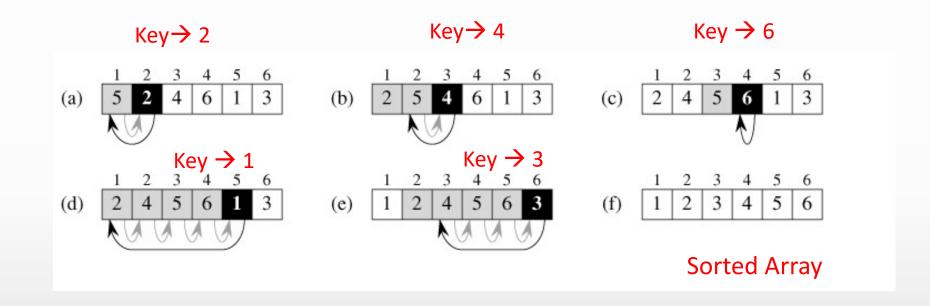
eg: sum = sum + A[i] is a one step in the CISC processor.

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



8

A[i+1] = key

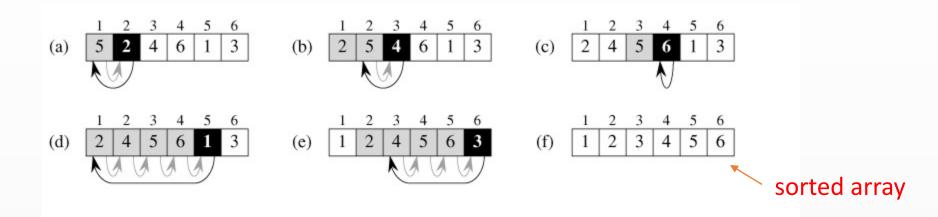


### Pseudocode for insertion sort.

#### **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 - 15 While i > 0 and A[i] > key 6 A[i+1] = A[i]i = i-1



### 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  $\rightarrow$  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:**

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

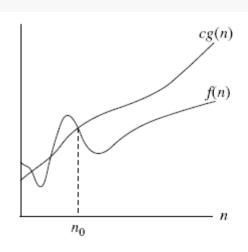
$$0 \le f(n) \le cg(n)$$
 for all  $n \ge n_o$ .

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.

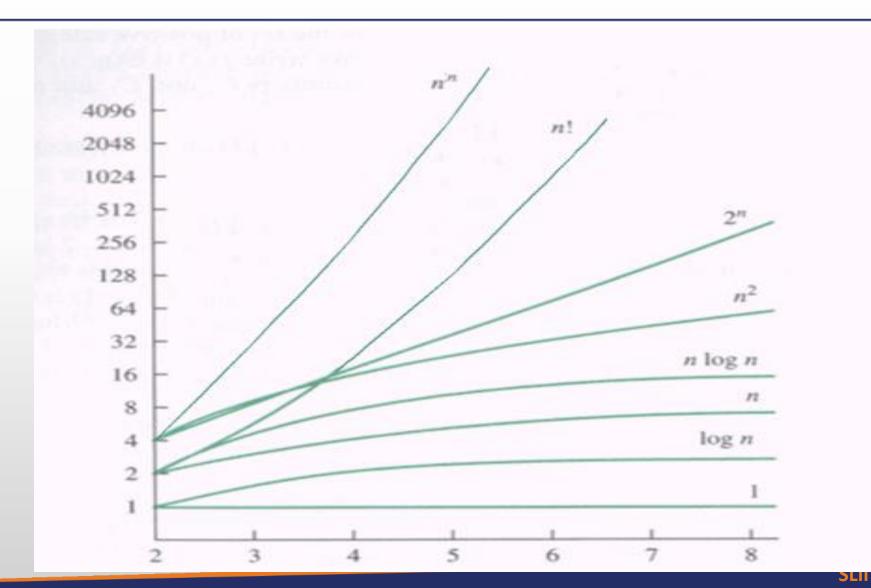
```
for i = 1 to n

for j = 1 to i

Print j
O(n^2)
```



### Graphs of functions





n	logn	n	nlog $n$	$n^2$	$n^3$	2 <sup>n</sup>
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 * 10 <sup>19</sup>
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 * 10 <sup>154</sup>
1024	10	1,024	10,240	1,048,576	1,073,741,824	1.79 * 10 <sup>308</sup>



# 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)



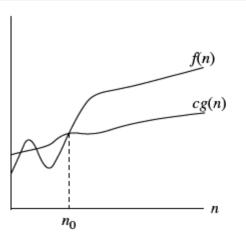
#### $\Omega$ - Notation

Provides the lower bound of the function.

#### **Definition:**

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

$$0 \le cg(n)$$



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

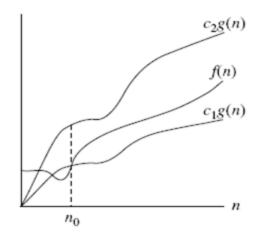


#### Θ - Notation

 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.