

# **DESIGN AND ANALYSIS OF ALGORITHMS**

**CS 4120/5120**

**THE EFFICIENCY OF ALGORITHMS**

# AGENDA

- Model of Implementation
- The complexity of an algorithm
  - Running time
  - Analyze running using **cost** and **time**
- Derive the running time function

# ANALYZING ALGORITHMS

- Analyzing an algorithm means *predicting the resources that the algorithm requires*.
  - Memory
  - Communication bandwidth
  - Computer hardware
  - **Computational time**
    - Algorithms will be *implemented by computer programs*.
      - The computational time (performance) can be affected by the above aspects

# CASE

- Consider computers A and B. Both computer executes one instruction per clock cycle of their respective CPUs.

Computer	Architecture	Clock cycle time
A	Reduced Instruction Set Computer (RISC)	80 ps
B	Complex Instruction Set Computer (CISC)	1000 ps

– Computer B's instruction set includes an instruction for sorting integers

- Run the same sorting program on the same input data on both computers.
- Which computer is likely to finish sorting first?

# IMPLEMENTATION TECHNOLOGY

## THE RAM MODEL

- A generic one-processor, **Random-Access Machine (RAM)** model of computation
- Certain constraints on the ISA apply

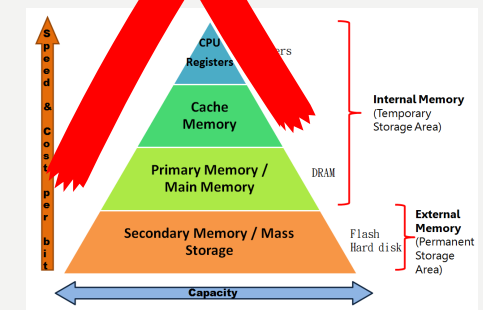
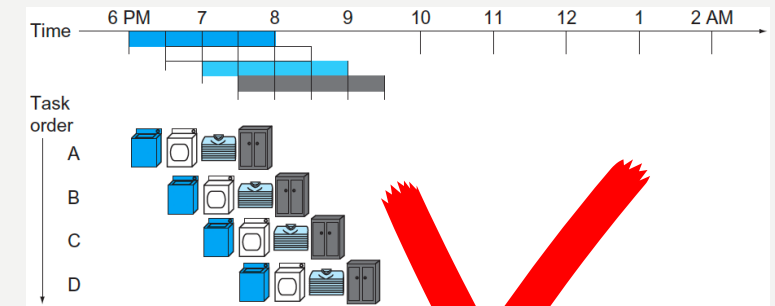
# RANDOM-ACCESS MACHINE (RAM)

## INSTRUCTION CONSTRAINTS

- The instruction set of the RAM model contains **instructions commonly found in real computers**.
  - Arithmetic
    - $+$ ,  $-$ ,  $\times$ ,  $\div$ , *mod*, *floor* ( $\lfloor \ \rfloor$ ), *ceiling* ( $\lceil \ \rceil$ )
  - Data movement
    - load, store, copy
  - Control
    - conditional and unconditional branches, subroutine call and return

# RANDOM-ACCESS MACHINE (RAM) INSTRUCTION EXECUTION

- The program instructions of the RAM model are executed one after another, with **no concurrent operations**.
- Each instruction takes a **CONSTANT amount of time**.
  - The computational time of exponentiation is specified in the next slide.
- **Do NOT model memory hierarchy**



# RANDOM-ACCESS MACHINE (RAM)

## EXPONENTIATION EXECUTION TIME

- In general, computing  $x^y$  when  $x$  and  $y$  are real numbers is **NOT a constant-time operation**.
  - Example:  $1.5^{2.1}$ ,  $(\sqrt{2})^{3.5}$ ,  $5^{0.5}$ ,  $1.5^y$ , where  $y \in R$ .  $R$  is the set of real numbers.
- However, computing  $2^k$  when  $k$  is a **small enough positive integer** is a constant-time operation.



# CONSTANT-TIME OPERATION PRACTICE

CONSIDER THE FOLLOWING OPERATIONS CARRIED OUT BY A RAM.

- Determine whether **each statement** is a constant-time op.
- Then, determine whether **each lettered item** is a constant-time op.

A. **int** x = 5;

B. y = a + b;

C. **for** i = 1 **to** n

A[i] = A[i] + 1

D. **for** i = 1 **to** 100

A[i] = A[i] + 1

E.  $n = 2^m$ , where m is an integer, and  $m \in (0, 10]$

F. **if**  $n > A.length$  //A is an array

G.  $n = 2^k$ , where k is a real number, and  $k \in (0, 10]$ .

H.  $n = m^k$

# CONSTANT-TIME OPERATION PRACTICE

- Consider the algorithm. Is each statement a constant-time operation?

```
INSERTION-SORT(A)
1  for j = 2 to A.length
2      key = A[j]
3      // Insert A[j] into the sorted sequence A[1..j - 1]
4      i = j - 1
5      while i > 0 and A[i] > key
6          A[i + 1] = A[i]
7          i = i - 1
8      A[i + 1] = key
```

# THE EFFICIENCY OF AN ALGORITHM

## COMPLEXITY

- The algorithm can solve the problem with limited resources.
  - Time complexity: the job could be done within *finite* time.
    - Also called the **running time** of an algorithm.
      - The **running time** (time complexity) of an algorithm, denoted by  $T(n)$ , is the number of primitive operations or “steps” executed
  - Space complexity: the amount of available memory is determinate.
    - Compared to time complexity, space complexity is given less consideration in this course when designing an algorithm.

# THE EFFICIENCY OF AN ALGORITHM

## RUNNING TIME

- In other words, **running time** of an algorithm is the sum of running times for each statement executed.

- Example

- Consider the code below

1	<b>for</b> $i = 1$ <b>to</b> $n$	Cost $c_1$	Time $n + 1$
2	$A[i]++$	$c_2$	$n$

- The running time of the code is  $T(n) = c_1 \cdot (n + 1) + c_2 \cdot n$

The computational cost of a **single** statement

The number of executions of a **single** statement

# RUNNING TIME PRACTICE

- Compute the running time of following (independent) code segments using the **cost-time** table.

A.	Cost	Time	Running Time
1 <b>for</b> $i = 2$ <b>to</b> $n$			$T(n) =$
2 $A[i]++$			

B.	Cost	Time	Running Time
1 <b>for</b> $j = 5$ <b>to</b> $n$			$T(n) =$
2 $i = j - 1$			
3 <b>while</b> $i > 0$			
4 $i--$			

# THE EFFICIENCY OF AN ALGORITHM

## ANALYZE INSERTION-SORT

- Derive the close-end form of the running time function  $T(n)$  of INSERTION-SORT
  - Use  $\sum$  and  $t_j$  to denote the executions of the **while**-loop at line 5 for a value of  $j$
- See next slide.

	Cost	Time
1 <b>for</b> $j = 2$ <b>to</b> $A.length$	$c_1$	$n$
2 $key = A[j]$	$c_2$	$n - 1$
3     // Insert $A[j]$ into the sorted sequence $A[1..j - 1]$	$c_3 = 0$	
4 $i = j - 1$	$c_4$	$n - 1$
5 <b>while</b> $i > 0$ <b>and</b> $A[i] > key$	$c_5$	$\sum t_j$
6 $A[i + 1] = A[i]$	$c_6$	$\sum (t_j - 1)$
7 $i = i - 1$	$c_7$	$\sum (t_j - 1)$
8 $A[i + 1] = key$	$c_8$	$n - 1$

# THE EFFICIENCY OF AN ALGORITHM

## ANALYZE INSERTION-SORT (CONT'D)

- Derive the close-end form of the running time function  $T(n)$  of INSERTION-SORT

$$T(n) = c_1n + c_2(n - 1) + c_4(n - 1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n - 1)$$

	<i>INSERTION-SORT</i> ( <i>A</i> )	Cost	Time
1	<b>for</b> <i>j</i> = 2 <b>to</b> <i>A.length</i>	$c_1$	$n$
2	$key = A[j]$	$c_2$	$n - 1$
3	// Insert $A[j]$ into the sorted sequence $A[1..j - 1]$	$c_3 = 0$	
4	$i = j - 1$	$c_4$	$n - 1$
5	<b>while</b> $i > 0$ <b>and</b> $A[i] > key$	$c_5$	$\sum t_j$
6	$A[i + 1] = A[i]$	$c_6$	$\sum (t_j - 1)$
7	$i = i - 1$	$c_7$	$\sum (t_j - 1)$
8	$A[i + 1] = key$	$c_8$	$n - 1$

# THE EFFICIENCY OF AN ALGORITHM

## SIMPLIFY THE FUNCTION

- Simplify the running time function

$$\begin{aligned}
 T(n) &= c_1n + c_2(n-1) + c_4(n-1) + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n (t_j - 1) + c_7 \sum_{j=2}^n (t_j - 1) + c_8(n-1) \\
 &= c_1n + c_2n - c_2 + c_4n - c_4 + c_5 \sum_{j=2}^n t_j + c_6 \sum_{j=2}^n t_j - c_6 \sum_{j=2}^n 1 + c_7 \sum_{j=2}^n t_j - c_7 \sum_{j=2}^n 1 + \\
 &\quad c_8n - c_8 \\
 &= (c_1 + c_2 + c_4 + c_8)n - (c_2 + c_4 + c_8) + (c_5 + c_6 + c_7) \sum_{j=2}^n t_j - (c_6 + c_7) \sum_{j=2}^n 1
 \end{aligned}$$



# THE EFFICIENCY OF AN ALGORITHM

## SIMPLIFY THE FUNCTION (CONT'D)

- Simplify the running time function

$$\begin{aligned}
 T(n) &= (c_1 + c_2 + c_4 + c_8)n - (c_2 + c_4 + c_8) + (c_5 + c_6 + c_7) \sum_{j=2}^n t_j - (c_6 + c_7) \underbrace{\sum_{j=2}^n 1}_{\substack{\parallel \\ (n-1)}} \\
 &= (c_1 + c_2 + c_4 + c_8)n - (c_2 + c_4 + c_8) + (c_5 + c_6 + c_7) \sum_{j=2}^n t_j - (c_6 + c_7)(n-1) \\
 &= (c_1 + c_2 + c_4 + c_8 - c_6 - c_7)n - (c_2 + c_4 + c_8 + c_6 + c_7) + (c_5 + c_6 + c_7) \sum_{j=2}^n t_j
 \end{aligned}$$

This term determines the running time is linear or quadratic.

# THE EFFICIENCY OF AN ALGORITHM

## BEST- VS WORST-CASE

- At this point, we have obtained the running time function in terms of  $n$  and  $t_j$ .

$$T(n) = (c_1 + c_2 + c_4 + c_8 - c_6 - c_7) n - (c_2 + c_4 + c_8 + c_6 + c_7) + (c_5 + c_6 + c_7) \sum_{j=2}^n t_j$$

- $t_j$  depends on the actual input array  $A$
- More specifically,  $t_j$  depends on  $j$  and the **relationship between each element of subarray  $A[1..j-1]$  and  $A[j]$** .

# THE EFFICIENCY OF AN ALGORITHM

## BEST- VS WORST-CASE (CONT'D)

		Cost	Time	
			Best-case	Worst-case
<ul style="list-style-type: none"> <li>Revisit the algorithm under best-case and worst-case scenarios               <ul style="list-style-type: none"> <li>Complete the missing cells                   <ul style="list-style-type: none"> <li>Best-case scenario <math>t_j =</math></li> <li>Worst-case scenario <math>t_j =</math></li> </ul> </li> </ul> </li> </ul>	<i>INSERTION-SORT(A)</i>			
	1 <b>for</b> $j = 2$ <b>to</b> $A.length$	$c_1$	$n$	$n - 1$
	2 $key = A[j]$	$c_2$	$n - 1$	$n - 1$
	3     // Insert $A[j]$ into the sorted sequence $A[1..j - 1]$	$c_3 = 0$	NA	NA
	4 $i = j - 1$	$c_4$	$n - 1$	$n - 1$
	5 <b>while</b> $i > 0$ <b>and</b> $A[i] > key$	$c_5$		
	6 $A[i + 1] = A[i]$	$c_6$		
	7 $i = i - 1$	$c_7$		
	8 $A[i + 1] = key$	$c_8$	$n - 1$	$n - 1$

# THE EFFICIENCY OF AN ALGORITHM

## RUNNING TIME OF INSERTION-SORT

- **Best-case scenario**
  - Input A is already **sorted** in the desired order
  - $T(n) = (c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8)$ 
    - In the form of  $an + b$
  - Running time is a **linear function**.

# THE EFFICIENCY OF AN ALGORITHM

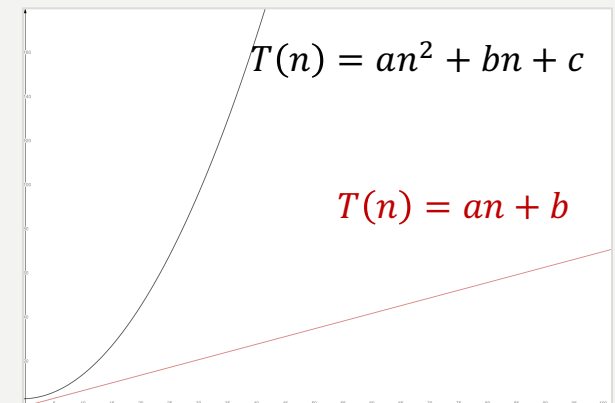
## RUNNING TIME OF INSERTION-SORT

- **Worst-case scenario**
  - Input A is **reverse** sorted .
  - $T(n) = \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2}\right)n^2 + \left(c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8\right)n - (c_2 + c_4 + c_5 + c_8)$ 
    - In the form of  $an^2 + bn + c$
  - Running time is a **quadratic function**.

# THE EFFICIENCY OF AN ALGORITHM

## RUNNING TIME OF INSERTION-SORT

- Compare the running time functions
  - Best-case:  $T(n) = (c_1 + c_2 + c_4 + c_5 + c_8)n - (c_2 + c_4 + c_5 + c_8)$
  - Worst-case:  $T(n) = \left(\frac{c_5}{2} + \frac{c_6}{2} + \frac{c_7}{2}\right)n^2 + \left(c_1 + c_2 + c_4 + \frac{c_5}{2} - \frac{c_6}{2} - \frac{c_7}{2} + c_8\right)n - (c_2 + c_4 + c_5 + c_8)$



# **NEXT UP**

# **GROWTH FUNCTIONS**

# REFERENCES