

Chapter 3. Algorithm

- ▶ 1. Algorithm definition
 - pseudocode
 - examples
 - Recursive algorithm
- ▶ 2. Analysis of Algorithm
 - Time Complexity
 - Order notation



1. Introduction

- ▶ An *algorithm* is a step-by-step method for solving some problem
- a finite set of instructions with the following characteristics:
 - ❖ **Precision:** steps are precisely stated
 - ❖ **Uniqueness:** The intermediate results of each step of execution are uniquely defined. They depend only on inputs and results of preceding steps.
 - ❖ **Finiteness:** the algorithm stops after finitely many steps
 - ❖ **Input/Output :** the algorithm receives input and produces output
 - ❖ **Generality:** the algorithm applies to various sets of inputs



Notation for algorithms

- methods: programming language (too detail),
flowchart, pseudocode,...
- Pseudocode: English like + PL style (looks computer language such as C++ or Pascal), free of details
easy to translate to PL.

► General form of PSEUDOCODE

BEGIN

{ statements }

END.



1.1 Pseudocode

1)Assignment statement: assigns value to a variable.

ex) Compute sum of two numbers, 1st and 2nd, And assigns result to SUM

BEGIN

INPUT first and second

sum \leftarrow first + second

END.

ex) Algorithm to find the largest of three numbers a, b, c:

1. input a, b, c

2. large= a ;; “copy the value of a into large”

3. If b > large then large= b

4. If c > large then large= c

5. return large.

Ex) find smallest of three numbers a,b,c,?

Pseudocode

2) **Control statements:** flow of control through the algorithm

a) sequence : List of statements to be executed as a single unit

Ex) procedure max(a,b,c)

begin

x:=a;

If b>x then x:= b

If c>x then x:= c

Return(x)

End max

b) conditional: if-then or if-then-else

- if p then action1 else action2

- if p then begin

action1, action2, ... actionN

endif

c) iterative (loop)

- FOR variable := initial value TO final value DO {Statements}

- WHILE (expression) DO { statements }

▶ - REPEAT {Statements} UNTIL (condition)

1.2 Algorithm examples

- 1) Find sum of first n odd numbers 2) test a positive integer is prime

Procedure find_odd

```
begin
  sum  $\leftarrow$  0
   $l \leftarrow 1$ 
  input n
  while  $l \leq n$  do
    begin
      sum  $\leftarrow$  sum +  $l$  ;
       $l \leftarrow l+2$ 
    end
  output sum
end
```

procedure is_prime(m)

```
  for  $i=2$  to  $m-1$  do
    if  $m \bmod i = 0$ , then return false
  return true
end is_prime
```

- 3) find a prime larger than a given integer

procedure large_prime

```
   $m=n+1$ 
  while not(is_prime(m)) do
     $m=m+1$ 
  return m
end large_prime
```



More examples

* Euclidean algorithm

finding the greatest common divisor between two integers.

Ex) GCD of 4 and 6 = 2

if a, b, q are integers, $b \neq 0$, satisfying $a = bq$, then

- ▶ b DIVIDES a ,
- ▶ writes $b \mid a$ (q = quotient, b =divisor of a)
(if b does not divide, then we write $b \nmid a$)

▶ Theorem:

If a is a nonnegative integer, b is a integer, and

$a = bq + r$, $0 \leq r < b$ then $\text{Gcd}(a, b) = \text{gcd}(b, r)$



GCD procedure

► Procedure gcd(a,b)

If $a < b$ then swap(a,b)

While $b \text{ NEQ } 0$ do

begin

divide a by b to obtain $a = bq + r$, ($0 \leq r < b$)

$a := b$

$b := r$

end

return(a)

End GCD



1.3 Recursive algorithms

- A *recursive procedure* is a procedure that invokes itself (divide-conquer)
- A *recursive algorithm* is an algorithm that contains a recursive procedure

(Ex: **factorial of n** is defined as, $n! = n(n-1)(n-2)\dots 3.2.1$)

► Ex) Procedure fact(n)
 If n=0 then Return (1)
 Return (n*fact(n-1))
End

Procedure iterative-fact (n)
 X=1;
 For I = 1 to n { x= i * x}
End

- Fibonacci sequence f_1, f_2, \dots defined recursively as follows:

$$f_1 = 1, \quad f_2 = 2$$

$$f_n = f_{n-1} + f_{n-2} \quad \text{for } n \geq 3$$

- First terms of the sequence are: 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, 144, 233, 377, 610, 987, 1597,...



Recursive algorithm example

- ▶ A Robot can take steps of 1meter or 2meters. An algorithm to calculate the number of ways the robot can walk n meters

Distance	sequence of steps	ways to walk
1	1	1
2	1,1 or 2	2
3	1,1,1 or 1,2 or 2,1	3
4	1,1,1,1 or 1,1,2 or 1,2,1 or 2,1,1 or 2,2	5

if $n > 2$, if robot takes 1meter step, then $n-1$ meters remain
if robot takes 2meter step, then $n-2$ meters remain

Algorithm:

```
Walk(n) {  
    If (n==1 v n==2) return n  
    Return walk(n-1) + walk(n-2)  
} * Fibonacci sequence
```

Recursive algorithm example (이진 탐색)

▶ 예제2 [이진 탐색] : 정수 searchnum이 배열 list에 있는지 검사

=> $list[0] \leq list[1] \leq \dots \leq list[n-1]$ /* 미리 정렬되어 있음 */

=> $list[i] = searchnum$ 인 경우 인덱스 i 를 반환, 없는 경우는 -1 반환

(초기값: $left = 0, right = n-1$; list의 중간 위치: $middle = (left + right) / 2$)

* $list[middle]$ 과 $searchnum$ 비교 시 3가지중 하나를 선택

1) $searchnum < list[middle]$: /* search again between left and middle-1 */


2) $searchnum = list[middle]$: /* middle을 반환 */

3) $searchnum > list[middle]$: /* search again between middle+1 and right */



이진탐색 알고리즘

```
int binarysearch(int list[], int num, int left, int right) {  
    while (left <= right) {  
        middle = (left + right) / 2;  
        if (num < list[middle])    right = middle - 1;  
        else if (num == list[middle])    return middle;  
        else left = middle + 1;  
    }  
    int binsearch(int list[], int num, int left, int right) {  
        if (left <= right) {  
            middle = (left + right) / 2;  
            if (num < list[middle]) binsearch(list, num, middle-1, right);  
            else if (num == list[middle]) return middle;  
            else binsearch(list, num, left, middle+1);  
        }  
        return -1;  
    }  
}
```



Algorithm 4.4.2: Computing n Factorial

```
1.  factorial( $n$ ) {  
2.    if ( $n == 0$ )  
3.      return 1  
4.    return  $n * \textit{factorial}(n - 1)$   
5.  }
```

Algorithm 4.4.6: Robot Walking

Input: n

Output: $walk(n)$

```
 $walk(n)$  {  
  if ( $n == 1 \vee n == 2$ )  
    return  $n$   
  return  $walk(n - 1) + walk(n - 2)$   
}
```

Algorithm 4.1.1: Finding the Maximum of Three Numbers

Input: a, b, c

Output: $large$ (the largest of a, b , and c)

```
1.   $max3(a, b, c)$  {  
2.     $large = a$   
    // if  $b$  is larger than  $large$ , update  $large$   
3.    if ( $b > large$ )  
4.       $large = b$   
    // if  $c$  is larger than  $large$ , update  $large$   
5.    if ( $c > large$ )  
6.       $large = c$   
7.    return  $large$   
8.  }
```

Algorithm 4.1.2: Finding the Maximum Value in a Sequence

Input: s, n

Output: $large$ (the largest value in the sequence s)

```
 $max(s, n)$  {  
     $large = s_1$   
    for  $i = 2$  to  $n$   
        if ( $s_i > large$ )  
             $large = s_i$   
    return  $large$   
}
```


Algorithm 4.2.1: Text Search

Input: p (indexed from 1 to m), m , t (indexed from 1 to n), n

Output: i

```
text_search( $p, m, t, n$ ) {  
  for  $i = 1$  to  $n - m + 1$  {  
     $j = 1$   
  
    //  $i$  is the index in  $t$  of the first character of the  
    // substring to compare with  $p$ , and  $j$  is the index in  $p$   
  
    // the while loop compares  $t_i \cdots t_{i+m-1}$  and  $p_1 \cdots p_m$   
    while ( $t_{i+j-1} == p_j$ ) {  
       $j = j + 1$   
      if ( $j > m$ )  
        return  $i$   
    }  
  }  
  return 0  
}
```

2. 알고리즘분석 (Analysis of algorithms)

- ▶ *Complexity*: Amount of time and/or space needed to execute algorithm.
- ▶ Complexity depends on many factors: data representation type, kind of computer, computer language used, etc.
- ▶ **Alternative method: - count basic operation/program steps**

ex) function sum(a: Elementlist; n: integer): real;

1 var s: real; I: integer; 0

2 s:= 0 1

3. for i:= 1 to n do n+1

4. s:= s+ a[i] n

5. sum:= s; 1

total steps: $2n+3$

ex) for i=1 to n do

for j=1 to n do

x=x+1

ex) while (n \geq 1) do {

for i=1 to n do

x=x+1

n= n/2 }

2.1 시간 복잡도 (Time complexity)

- ▶ Time complexity function, $T(n)$: when n is size of problem, $T(n)$ is the estimation of running time for the algorithm.

ex) If $t(n)$ is, $60n^2 + 5n + 1$

n	$t(n) = 60n^2 + 5n + 1$	$60n^2$	
10	6051	6,000	- for large $60n^2$ is approx. equal to $t(n)$
1000	60,005,001	60,000,000	- ignore constants, so $t(n)$ grows like n^2 as n increases
10000	6,000,050,001	6,000,000,000	

- idea is to replace $t(n) = 60n^2 + 5n + 1$ with simpler expression as n^2

* 알고리즘의 정확한 수행시간보다도, 알고리즘의 수행시간이 input N 에 따라 얼마나 증가하는가에 관심이 있음.

2.2 Order notation (차수표기법)

Compare the time complexities of two programs that computes same function and to predict the growth in run time as the instance characteristics changes.

1) Definition: (Big-Oh), $f(n) = O(g(n))$; $f(n)$ is AT MOST $g(n)$

- f of n is big oh of g of n , iff there exist positive constants C and n_0

$$\therefore \underline{|f(n)| \leq C|g(n)|}, \quad \underline{\forall n, \quad n \geq n_0}$$

$g(n)$ 은 $f(n)$ 의 상한선(upper bound).

ex) $3n+2 =$

ex) $100n+6=$

ex) $1000n^2+100n-6 =$

► THM: If f_1 is $O(g_1)$ and f_2 is $O(g_2)$ then

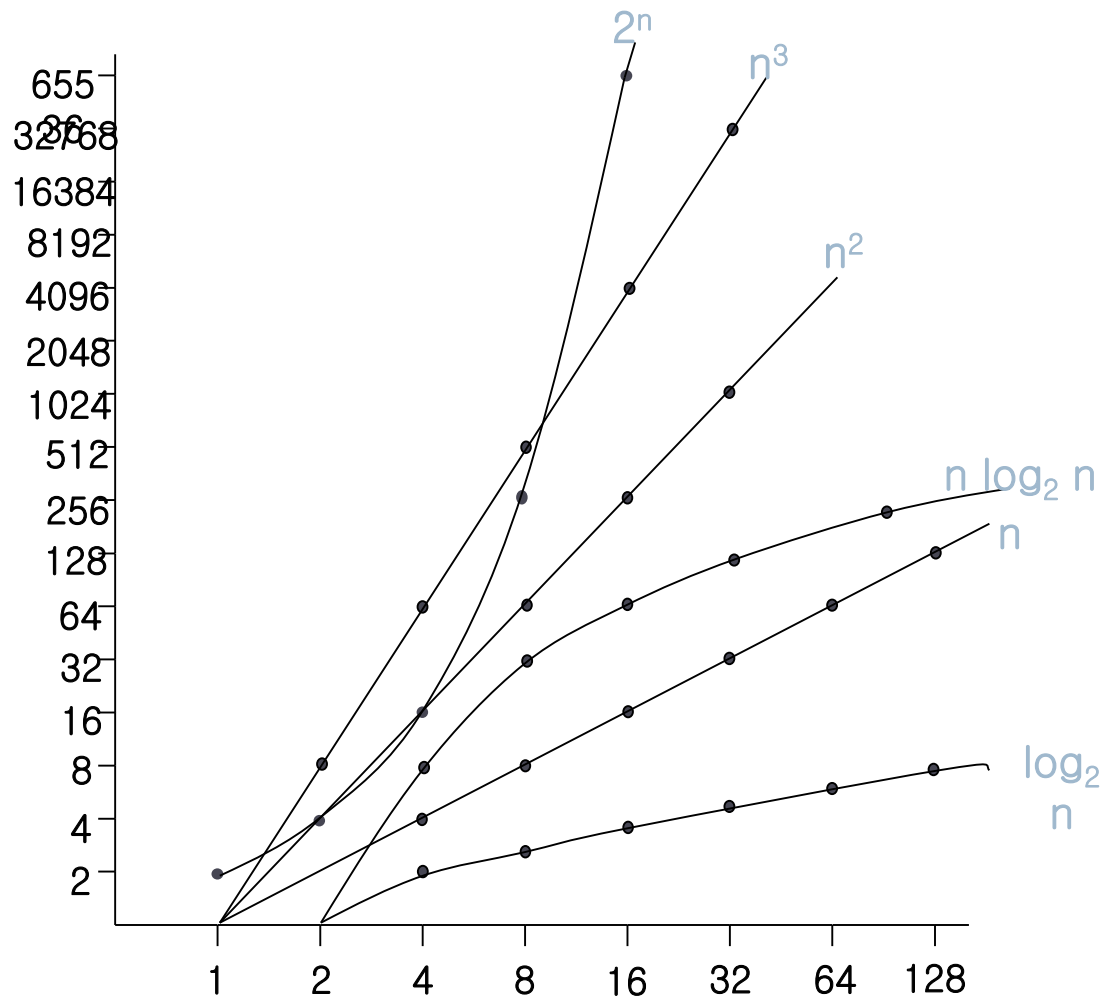
1) $f_1 f_2$ is $O(g_1 g_2)$

2) $f_1 + f_2$ is $O(\max\{g_1, g_2\})$



알고리즘의 난이도/복잡도 (Complexity Classes)

$O(1) < O(\log n) < O(n) < O(n \log n) < O(n^2) < O(n^3) \dots O(n^k) < O(2^n)$



▶ ▶ 그림 10-1 입력 크기에 따른 각 함수의 증가 비율

Order notation (차수표기법) – con't

2) Definition: (OMEGA), $f(n) = \Omega(g(n))$; $f(n)$ is AT LEAST $g(n)$

- f of n is omega of g of n , iff \exists positive constants C and n_0

$\therefore \underline{|f(n)| \geq C_1|g(n)|}, \underline{\forall n, n \geq n_0}, \quad g(n) \text{ 은 } f(n) \text{ 의 upper bound.}$

ex) $3n+2 =$

ex) $100n+6 =$

ex) $1000n^2+100n-6 =$

3) Definition: (THETA), $f(n) = \Theta(g(n))$;

iff there exist positive constants C_1, C_2 , and n_0 such that

$$C_1g(n) \leq f(n) \leq C_2g(n), \text{ for all } n, \quad n \geq n_0$$

ex) $f(n) = 2n + 3 \log n$

ex) $f(n) = 60n^2 + 5n + 1$

ex) $f(n) = a_m n^m + \dots + a_1 n + a_0$, and $a_m > 0$, then $f(n) = \Theta(n^m)$



Examples

ex) $1+2+\dots+n$ $\leq n+n+\dots+n = n \cdot n = n^2$ for all $n \geq 1$ (1)

\Rightarrow for upper bound: $1+2+\dots+n = O(n^2)$

\Rightarrow for lower bound, $1+2+\dots+n \geq 1+1+\dots+1 = n \Rightarrow 1+2+\dots+n = \Omega(n)$

but we cannot deduce Θ -estimate since, the lower bound and upper bound are not equal.

$$1+2+\dots+n \geq \lceil n/2 \rceil + \lceil n/2 \rceil + \dots + \lceil n/2 \rceil =$$

$$\lceil (n+1)/2 \rceil \lceil n/2 \rceil \geq (n/2)(n/2) = n^2/4 ; \text{ half of (1)}$$

► So we conclude that $1+2+\dots+n = \Omega(n^2)$ and $1+2+\dots+n = O(n^2)$.

ex) $j := n$

while $j \geq 1$ do

for $i := 1$ to j do { $x := x+1$; $j := \lfloor j/2 \rfloor$ }

end

