# Chapter 1: Algorithms: Efficiency, Analysis, and Order

## Contents

- 1.1 Algorithms
- 1.2 The Importance of Developing Efficient Algorithms
  - 1.2.1 Sequential Search vs. Binary Search
  - 1.2.2 The Fibonacci Sequence
- 1.3 Analysis of Algorithms
  - 1.3.1 Time Complexity Analysis
  - 1.3.2 Applying the Theory
- 1.4 Order
  - 1.4.1 An Intuitive Introduction to Order
  - 1.4.2 A Rigorous Introduction to Order

## 1.1 Algorithms

#### Problem

- **Ex.** 1.1: *Sorting* Sort the list *S* of *n* numbers in nondecreasing order. The answer is the numbers in sorted sequence.
- Ex. 1.2: Searching Determine whether the number x is in the list S of n numbers. The answer is yes if x is in S and no if it is not.
- **Ex**) Partition Decide whether a given multiset  $A = \{a_1, ..., a_n\}$  of n positive integers has a partition P such that  $\sum_{i \in P} a_i = \sum_{i \notin P} a_i$

### Parameters, Instance, and Solution

- Problem may contain Parameters
  - **Ex.** 1.1 Sorting : S, n
  - **Ex.** 1.2 Searching: S, n, x
  - $\blacksquare$  Ex) Partition: A, n
- Instance of a problem: each specific assignment of values to parameters
  - **Ex. 1.3: Instance of** *Sorting* 
    - S = [10, 7, 11, 5, 13, 8] and n = 6
  - **Ex. 1.4: Instance of** *Searching* 
    - S = [10, 7, 11, 5, 13, 8], n = 6, and x = 5
  - Instance of *Partition* 
    - $A = \{10, 7, 11, 5, 13, 8\}$  and  $n = 6 \rightarrow \text{sol} = \text{`no'}$
    - $A = \{10, 7, 11, 5, 15, 8\}$  and  $n = 6 \rightarrow \text{sol} = \text{`yes'}$
- Solution

## Algorithm

- Step by step procedure for producing the solution to each instance
- Def.(S. Sahni): An algorithm is a finite set of instructions that accomplish a particular task.
  - Input zero or more
  - Output zero or more
  - Definiteness each instruction is clear and unambiguous
    - Ex) add 6 or 7 to x (X)
  - Finiteness: must terminate after a finite number of steps.
    - cf) procedure
    - OS (X)



- Effectiveness: each instruction must be very basic so that it can be carried out by a person using pencil and paper. It also must be feasible.
  - Integer arithmetic (O)
  - Real arithmetic (X) decimal expansion might be infinitely long
- Pseudocode: C++ like
- Program: expression of an algorithm in a PL

## 1.2 Importance of Developing Efficient Algorithms

## ■ 1.2.1 Sequential search vs. Binary search

- Algorithm 1.1 vs. Algorithm 1.5
- # of comparisons (worst case)
  - n vs.  $\log_2 n + 1$
- See Table 1.1

Table 1.1	The number of comparisons done by Sequential Search
and Binary	Search when $x$ is larger than all the array items

Array Size	Number of Comparisons by Sequential Search	Number of Comparisons by Binary Search	
128	128	8	
1,024	1,024	11	
1,048,576	1,048,576	21	
4,294,967,296	4,294,967,296	33	



- **Problem:** Is the key x in the array S of n keys?
- Input (parameters): positive integer n, array of keys S indexed from 1 to n, and a key x.
- Output: *location*, the location of x in S (0 if x is not in S)

```
void seqsearch (
                     int n,
                     const keytype S[],
                     keytype x,
                     index& location)
  location = 1;
  while (location <= n && S[location] != x)
    location ++;
  if (location > n)
    location = 0;
```



### Algorithm 1.5 Binary Search (1/2)

- Problem: Determine whether x in the sorted array S of n keys.
- Inputs: positive integer n, sorted (nondecreasing order) array of keys S indexed from 1 to n, and a key x.
- Outputs: location, the location of x in S (0 if x is not in S)

### Algorithm 1.5 Binary Search (2/2)

```
void binsearch (
                      int n,
                       const keytype S[],
                       keytype x,
                       index& location)
  index low, high, mid;
  low = 1; high = n;
  location = 0;
  while (low \leq high && location == 0) {
    mid = \lfloor (low + high) / 2 \rfloor;
     if (x == S[mid])
       location = mid;
     else if (x < S[mid])
       high = mid - 1
     else
       low = mid + 1;
```



## 1.2.2 Fibonacci Sequence

$$f_0 = 0,$$
  
 $f_1 = 1,$   
 $f_n = f_{n-1} + f_{n-2} \quad (n \ge 2)$ 

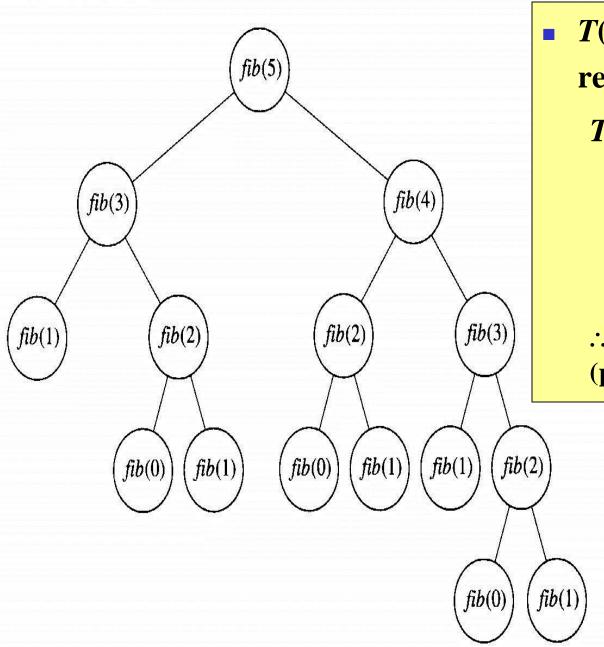


## Algorithm 1.6 *n*–th Fibonacci Term (Recursive)

- **Problem:** Determine the *n*-th term in the Fibonacci Sequence.
- **Inputs**: a nonnegative integer *n*.
- **Outputs:** *fib*, the *n*-th term of the Fibonacci Sequence.

```
int fib (int n)  \{ \\  if \ (n <= 1) \\      return \ n; \\      f_0 = 0, \ f_1 = 1, \\      f_n = f_{n-1} + f_{n-2} \ (n \ge 2) \\  else \\      return \ fib (n-1) + fib (n-2); \\ \}
```

Figure 1.2 The recursion tree corresponding to Algorithm 1.6 when computing the fifth Fibonacci term.



• T(n): # of terms in the recursion tree for n.

$$T(n) > 2 \times T(n-2)$$
  
> 2 x 2 x  $T(n-4)$ 

$$> 2 \times 2 \times ... \times 2 \times T(0)$$
n/2 terms

$$T(n) > 2^{n/2}$$
(proof by induction in Th. 1.1)



- **Problem:** Determine the *n*-th term in the Fibonacci Sequence.
- <u>Inputs</u>: a nonnegative integer n.
- **Outputs:** *fib2*, the *n*-th term in the Fibonacci Sequence.

```
int fib2 (int n)
  index i;
  int f[0..n];
  f[0] = 0;
  if (n > 0) {
     f[1] = 1;
     for (i = 2; i \le n; i++)
       f[i] = f[i-1] + f[i-2];
     return f[n];
```

dynamic programming:compute (n+1) terms

Table 1.2 A comparison of Algorithms 1.6 and 1.7

n	n + 1	2 <sup>n/2</sup>	Execution Time Using Algorithm 1.7	Lower Bound on Execution Time Using Algorithm 1.6
40	41	1,048,576	41 ns*	1048 μs <sup>†</sup>
60	61	$1.1 \times 10^{9}$	61 ns	1 s
80	81	$1.1 \times 10^{12}$	81 ns	18 min
100	101	$1.1 \times 10^{15}$	101 ns	13 days
120	121	$1.2 \times 10^{18}$	121 ns	36 years
160	161	$1.2 \times 10^{24}$	161 ns	$3.8 \times 10^7$ years
200	201	$1.3 \times 10^{30}$	201 ns	$4 \times 10^{13}$ years

<sup>\*1</sup> ns =  $10^{-9}$  second.

 $<sup>^{\</sup>dagger}1 \ \mu s = 10^{-6} \text{ second.}$ 



## 1.3 Analysis of Algorithms

- Efficiency
  - Space complexity: memory
  - Time complexity: execution time



- Want a measure independent of
  - Computer
  - Programming language
  - Programmer
  - Complex details of algorithms (pointer setting, incrementing of loop indices)
- Not want # of CPU cycles or instructions
  - Ex) binary search is more efficient than sequential search
    - # of comparisons:  $\log n < n$
- Algorithm's efficiency: # of basic operations executed as a function of input size



- Problem: Add all the numbers in the array S of n numbers.
- Inputs: positive integer n, array of numbers S indexed from 1 to n.
- Outputs: sum, the sum of the numbers in S.

```
number\ sum\ (int\ n,\ const\ number\ S[\ ]) \{ \\ index\ i; \\ number\ result; \\ result = 0; \\ for\ (i = 1;\ i <= n\ ;\ i++) \\ result = result + S[i]; \\ return\ result; \\ \}
```



- Problem: Sort n keys in nondecreasing order.
- Inputs: positive integer n, array of keys S indexed from 1 to n.
- Outputs: the array S containing the keys in nondecreasing order.

```
void exchangesort (int n, keytype S[ ]) {
   index i, j;
   for (i = 1; i <= n -1; i++)
      for (j = i+1; j <= n; j++)
      if (S[j] < S[i])
      exchange S[i] and S[j];
}</pre>
```



- Problem: Determine the product of two n x n matrices.
- Inputs: a positive integer n, 2D arrays of numbers A and B, each of which has both its rows and columns indexed from 1 to n.
- Outputs: a 2D array of numbers C, which has both its rows and columns indexed from 1 to n, containing the product of A and B.

```
void matrixmult (
                           int n,
                           const number A[][],
                           const number B[][],
                                  number C[ ] [ ])
  index i, j, k;
  for (i = 1; i \le n; i++)
     for (j = 1; j \le n; j++) {
        C[i][j] = 0;
                                                           /* b.o. \rightarrow e.t. = a
        for (k = 1; k <= n; k++)
          C[i][j] = C[i][j] + A[i][k] * B[k][j];
                                                           /* b.o. \rightarrow e.t. = b
```



### Input size and Basic operation

#### Input size

- Sequential search, binary search, add array members, exchange sort: array S of n keys
- Matrix multiplication: n, # of rows and columns
- Graph: n, e, # of nodes and edges
- Fibonacci number:  $\lfloor \log n \rfloor + 1$ , # of binary digits to encode n (n is input not input size)

#### Basic operation

- Single instruction or group of instructions
- **Execution time is independent of** n
- Ex) search: comparison

## Example

#### for (i = 1; i <= n; i++) for (j = 1; j <= n; j++) { $C[i][j] = 0; /* b.o. \rightarrow e.t. = a$ for (k = 1; k <= n; k++) C[i][j] = C[i][j] + A[i][k] \* B[k][j]; $/* b.o. \rightarrow e.t. = b$ }

#### Ex) matrix multiplication

- Execution time → see Algorithm 1.4
- Discussion

(1) 
$$a \cdot n^2 + b \cdot n^3$$

$$n = 10 \implies a \cdot 10^2 + b \cdot 10^3 \approx b \cdot 10^3$$

$$n = 100 \Rightarrow a \cdot 10^4 + b \cdot 10^6 \approx b \cdot 10^6$$

- (2) Ignore the time for incrementing loop indices
- (3) No time difference
  - temp = A[i][k] \* B[k][j];
  - C[i][j] = C[i][j] + temp



- **Determination of how many times the basic operations** is done for each value of the input size.
- In some cases, depends not only the input size but also on the *input value* 
  - Ex) sequential search

• Best case 
$$B(n) = 1$$

• Worst case 
$$W(n) = n$$

• Worst case 
$$W(n) = n$$
  
• Average case  $A(n) = (n+1)/2 \leftarrow \sum_{k=1}^{n} (k \times \frac{1}{n})$ 

Ex) array add

• Every case 
$$T(n) = n$$

Ex) Exchange sort

• 
$$T(n) = (n-1) + (n-2) + ... + 1 = (n-1)n/2$$

• Ex) matrix multiplication  $T(n) = n^3$ 

If 
$$T(n)$$
 exists  $T(n) = W(n) = A(n) = B(n)$   
If not  $W(n), A(n)$ 



- Fixed part that is independent of I/O characteristics
  - Instruction (code) space
  - Simple variable x = 3
  - Constants
  - Fixed size component variables (A[10], ...)
- Variable part
  - Variables depending on input size

 $\mathbf{E}\mathbf{x}: \mathbf{S}[\mathbf{n}], \mathbf{A}[\mathbf{n}][\mathbf{n}]$ 

■ Recursion stack (formal parameters, local variables, return address): space  $\geq 3(n+1)$  words, n is depth of recursion

Ex: Fibonacci number (Alg. 1.6): proportional to n