

# 4. Algorithms

## 4.1 Introduction

**Definition 1** An **algorithm** is a finite set of instructions for performing a computation or solving a problem.

*precision*

*steps are precisely stated*

*uniqueness*

*result of each step of execution are uniquely defined*

*depends on input and result of preceding step*

*determinancy vs. nondeterminancy*

*finiteness*

*the algorithm stops after finitely many steps*

***Dijkstra's mini language***

$$\begin{aligned}
 S ::= & x_1, \dots, x_n := E_1, \dots, E_n \mid S; S \mid \mathit{skip} \mid \mathit{abort} \\
 & \mid \mathit{if} B_1 \rightarrow SL_1 \mid B_2 \rightarrow SL_2 \mid \dots \mid B_n \rightarrow SL_n \mathit{fi} \\
 & \mid \mathit{do} B_1 \rightarrow SL_1 \mid B_2 \rightarrow SL_2 \mid \dots \mid B_n \rightarrow SL_n \mathit{od}
 \end{aligned}$$
***concurrent assignment***

$$x, y := y, x$$
***sequence of statements(S)***

$$S_1; S_2$$
***Two bases******skip******abort***

**Alternatervative construct**

$$\mathbf{if } B_1 \rightarrow SL_1 \mid B_2 \rightarrow SL_2 \mid \dots \mid B_n \rightarrow SL_n \mathbf{fi}$$

**Exa. Find maximum of two elements,  $x$  and  $y$**

$$\mathbf{if } x \geq y \mathbf{ then } m := x \mathbf{ else } m := y \mathbf{ fi}$$

$$\mathbf{if } x > y \mathbf{ then } m := x \mathbf{ else } m := y \mathbf{ fi}$$

$$\mathbf{if } x \geq y \rightarrow m := x \mid x \leq y \rightarrow m := y \mathbf{ fi}$$

$$\mathbf{if } x \geq y \rightarrow m := x$$

$$\mid x < y \rightarrow m := y$$

$$\mathbf{fi}$$

$$\mathbf{if } x > y \rightarrow m := x$$

$$\mid x \leq y \rightarrow m := y$$

$$\mathbf{fi}$$

$$\mathbf{if } x \geq y \rightarrow m := x$$

$$\mid x \leq y \rightarrow m := y$$

$$\mathbf{fi}$$

$$(m=x \vee m=y) \wedge m \geq x \wedge m \geq y$$

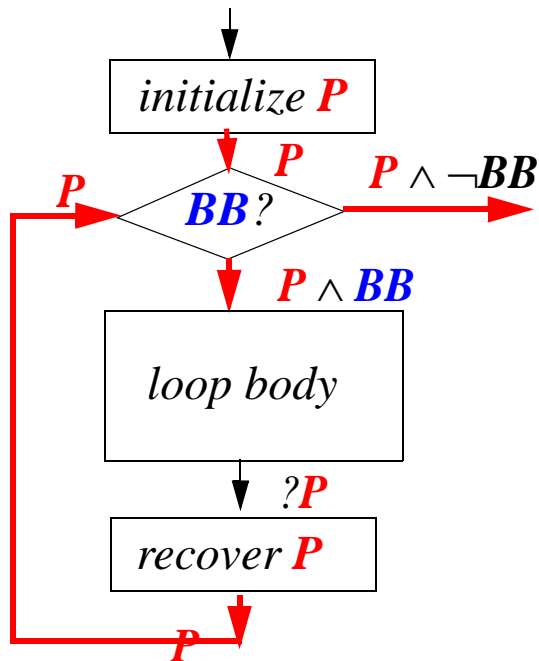
**nondeterminancy**

**$\mathbf{if } \mathbf{fi} \equiv \mathbf{abort}$**

**$\mathbf{do } \mathbf{od} \equiv \mathbf{skip}$**

### Repeatative construct

*do*  $B_1 \rightarrow SL_1$  '|' ... '|'  $B_n \rightarrow SL_n$  *od*



Let's assume that loop invariance is  $P$ .

and  $BB = B_1 \vee B_2 \vee \dots \vee B_n$

$$\begin{aligned} \therefore \neg BB &= \neg(B_1 \vee B_2 \vee \dots \vee B_n) \\ &= \neg B_1 \wedge \neg B_2 \wedge \dots \wedge \neg B_n \end{aligned}$$

### Loop invariance $P$

Loop invariance  $P$  should be **initialized before** the loop,

Loop invariance  $P$  must be **recovered in** the loop, if needed, and

Loop invariance  $P$  is still **valid after** (termination of) the loop.

$P \wedge \neg BB$  is the **final** condition that you want **after the loop!**

$a, b, c, d := A, B, C, D;$

**do**  $a > b \rightarrow a, b := b, a$

|  $b > c \rightarrow b, c := c, b$

|  $c > d \rightarrow c, d := d, c$

**od**

$$\begin{aligned} \neg \mathbf{BB} &= \neg((a > b) \vee (b > c) \vee (c > d)) = (a \leq b) \wedge (b \leq c) \wedge (c \leq d) \\ &= a \leq b \leq c \leq d. \end{aligned}$$

What is the loop invariance of the above do-od loop?

$\mathbf{P} = a, b, c, \text{ and } d \text{ are permutation of } A, B, C, \text{ and } D.$

Loop invariance  $\mathbf{P}$  should be **strong** enough to get **final result** with and  $\neg \mathbf{BB}$ ;

whereas **weak** enough

to be easily **initialized** and be **still valid** inside the loop.

$\mathbf{P} \wedge \neg \mathbf{BB}$  is the **final condition** that you want.

**Alg. 4.1.2 Finding the maximum value in a sequence**

**function**  $\text{max}(a_1, a_2, \dots, a_n: \text{numbers}): \text{number}$       $n \geq 1$   
 $\text{max} := a_1;$       $\text{max} = \text{max}(a_1)$   
**for**  $i := 2$  **to**  $n$  **do**  
   **if**  $\text{max} \leq a_i \rightarrow \text{max} := a_i$  /  $\text{max} \geq a_i \rightarrow$  **skip** **fi**      $\text{max} = \text{max}(a_1, \dots, a_i)$   
   **od**      $\text{max} = \text{max}(a_1, \dots, a_n)$

*Dijkstra's mini language*

$\text{max}, i := a_1, 2;$       $\text{max} = \text{max}(a_1, \dots, a_{i-1}) \wedge (i=2) \equiv \text{max} = \text{max}(a_1) \equiv \mathbf{T}.$   
**do**  $i \leq n \rightarrow$   
   **if**  $\text{max} \leq a_i \rightarrow \text{max} := a_i$  /  $\text{max} \geq a_i \rightarrow$  **skip**  
   **fi**;      $\text{max} = \text{max}(a_1, \dots, a_i)$   
    $i := i+1$       $\text{max} = \text{max}(a_1, \dots, a_{i-1})$   
**od**      $\text{max} = \text{max}(a_1, \dots, a_{i-1}) \wedge (i=n+1) \equiv \text{max}(a_1, \dots, a_n)$

## 4.2 Examples of Algorithms

**Alg. 4.2.1** The linear search algorithm

function *linear search*( $s_1, s_2, \dots, s_n$ : numbers with  $n \geq 1$ ):  $\mathbf{N}_{0,n}$ ;

$i := 1$ ; **do** ( $i \leq n \wedge x \neq s_i$ )  $\rightarrow i := i+1$  **od**;

$$\mathbf{P} \wedge \neg \mathbf{BB} = (1 \leq \forall k < i: x \neq s_k) \wedge ((i \leq n \wedge x = s_i) \vee (i = n+1))$$

**if**  $i \leq n \rightarrow loc := i$  |  $i = n+1 \rightarrow loc = 0$  **fi**;

$$((1 \leq \forall k < loc \leq n: x \neq s_k) \wedge (x = s_{loc})) \vee ((loc = 0) \wedge (1 \leq \forall k \leq n: x \neq s_k))$$

**return**  $loc$

Another implementation for the *linear search*

( $i, loc := 1, 0$ );  $(n \geq 0)$

**do** ( $(i \leq n) \wedge (loc = 0)$ )  $\rightarrow$

**if**  $x = s_i \rightarrow loc := i$  |  $x \neq s_i \rightarrow i := i+1$  **fi**

**od**

$$((1 \leq \forall k < loc \leq n: x \neq s_k) \wedge (x = s_{loc})) \vee ((loc = 0) \wedge (1 \leq \forall k \leq n: x \neq s_k))$$

**Alg. 4.2.1'** *The binary search algorithm*

procedure **binary search**( $x$ : number;  $s_1, s_2, \dots, s_n$ : **increasing numbers**)

$i, j, loc := 1, n, 0$ ;

**do**  $((i \leq j) \wedge (loc = 0)) \rightarrow$

$m := \lfloor (i+j)/2 \rfloor$ ;

**if**  $x > s_m \rightarrow i := m+1$

|  $x = s_m \rightarrow loc := m$

|  $x < s_m \rightarrow j := m-1$

**fi**

**od**

$(1 \leq \forall k < n: s_k \leq s_{k+1}) \wedge (n \geq 0)$  is the given condition (constant)

$(loc \leq n) \Rightarrow (x = a_{loc}) \vee (loc = 0) \Rightarrow (1 \leq \forall k \leq n: x \neq a_k)$



procedure **bubble sort**( $s_1, s_2, \dots, s_n$ : numbers with  $n \geq 2$ )

**for**  $i := 1$  to  $n-1$  **do**

$P1 \equiv (s_j = \max(s_1, s_2, \dots, s_j))$ . *init P1*:  $P1 \wedge (j=1) \equiv (s_1 = \max(s_1)) \equiv \mathbf{T}$ .

**for**  $j := 1$  to  $n-i$  **do**  $\rightarrow$

**if**  $s_j \geq s_{j+1} \rightarrow s_j, s_{j+1} := s_{j+1}, s_j$

|  $s_j \leq s_{j+1} \rightarrow \mathit{skip}$

**fi**

$P1 \equiv (s_j = \max(s_1, s_2, \dots, s_j))$ .

**od**

$P2 \equiv P1 \wedge (j=n-i+1) \equiv (s_{n-i+1} = \max(s_1, s_2, \dots, s_{n-i+1}))$

$P2 \wedge (i=1) \equiv (s_n = \max(s_1, s_2, \dots, s_n))$      *initialize P2*

**od**

$(1 \leq \forall i \leq n: P2) \equiv 1 \leq \forall i \leq n: (s_i = \max(s_1, s_2, \dots, s_i))$

$\equiv s_1 \leq s_2 \leq \dots \leq s_n$ .     *sorted in increasing order*

$$i = 1, j = 1 \quad P1 = (s_1 = \max(s_1)) \equiv \mathbf{T. \textit{init. of } P1(\textit{before loop})}$$

$$, j = 2 \quad P1 = (s_2 = \max(s_1, s_2)).$$

...

$$i = 1, j = n - i + 1 = n \quad P1 = (s_n = \max(s_1, s_2, \dots, s_n)) = P2. (n-1)\text{-comp.}$$

$$i = 2, j = 1 \quad P1 = (s_1 = \max(s_1)).$$

...

$$i = 2, j = n - 2 + 1 = n - 1 \quad P1 = (s_{n-1} = \max(s_1, s_2, \dots, s_{n-1})). (n-2)\text{-com.}$$

...

$$i = n - 2, j = n - (n - 2) + 1 = 3 \quad P1 = (s_3 = \max(s_1, s_2, s_3))$$

$$i = n - 1, j = 1 \quad P1 = (s_1 = \max(s_1)).$$

$$, j = n - i (= n - 1) + 1 = 2 \quad P1 = (s_2 = \max(s_1, s_2)) = P2. \quad 1 \textit{ comp.}$$

$$1 \leq \forall k \leq n: s_k = \max(s_1, s_2, \dots, s_k) \equiv s_1 \leq s_2 \leq \dots \leq s_n. \quad n(n-1)/2 \textit{ comp.}$$

**Greedy algorithm****Optimization problem**

*to find a solution to the given problem that*

*either **minimizes** or **maximizes** the value of some parameters*

**(global) optimal**

*to select the best choice considering all sequences of steps*

**local optimal (greedy algorithm)**

*to select the best choice at each step*

**Algorithm 6 Greedy Change-Making Algorithm**

*procedure change( $n_1, n_2, \dots, n_r$ : number of coins are  $r$  and*

*$c_1 > c_2 > \dots > c_r$  is the value for each coin $_i$ ;*

*$n$ : positive integer is the amount that you want to make change)*

*$n_1, n_2, \dots, n_r := 0, 0, \dots, 0; i := 1;$*

***do ( $i \leq r$ )  $\rightarrow$  do ( $n \geq c_i$ )  $\rightarrow n_i := n_i + 1; n := n - c_i$  od;  $i := i + 1$  od***

## 4.3 Analysis of Algorithms

### Big-O Notation

**Definition 1** Let  $f: \mathbf{N} \rightarrow \mathbf{R}^+$ .

We say  $f(n)$  is  $O(g(n))$ , if  $\exists C, k > 0 . \exists . \forall n > k: |f(n)| \leq C|g(n)|$ .

We also write  $f(n) = O(g(n))$  or even  $f(n) \in O(g(n))$ .

The constant  $C$  and  $k$  are called the **witnesses** of ...  $f(n)$  is  $O(g(n))$ .

Assume  $C$  and  $k$  are **smallest** witnesses, Then  $\forall C', k' . \exists . C < C', n < n',$   
 $C'$  and  $k'$  are also **witnesses**

$g(n)$  is **faster or equal(not slower)** growing than  $f(n)$   
 when  $n$  becomes **large enough**  
 and **ignoring** multiplying some **constant  $C$** .

**Theorem 1** Let  $f(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$ . Then  $f(x)$  is  $O(x^n)$ .  
*proof easy with witness  $k=1$  and  $C = \sum |a_i|$ .*

*Example 6*  $n!$

$n! \leq n^n$ .  $n!$  is  $O(n^n)$  with witness  $k=1$  and  $C=1$ .

$\log n! \leq \log n^n = n \log n$

$1, \log n, n, n \log n, n^2, \dots, 2^n, n!$

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

**Theorem 2, 3** Let  $f_1 \in O(g_1)$ ,  $f_2(x) \in O(g_2)$ . Then

$$f_1+f_2 \in O(g_1+g_2) = O(\max(|g_1|, |g_2|)) \text{ and}$$

$$f_1f_2 \in O(g_1g_2).$$

**Colollary 1** Let  $f_1, f_2 \in O(g)$ . Then  $f_1+f_2 \in O(g)$ .

## Big-Omega and Big-Theta Notation

**Definition 2**  $\Omega(g) = \{f: \mathbf{N} \rightarrow \mathbf{R}^+ \mid \exists C, k > 0 \forall n > k: |f(n)| \geq C|g(n)|\}$

If  $f \in \Omega(g)$ ,  $f$  grows at least order of  $g$ .

**Def. 3**  $\Theta(g) = \{f: \mathbf{N} \rightarrow \mathbf{R}^+ \mid \exists C_1, C_2, k > 0, \forall n > k: C_1|g(n)| \leq |f(n)| \leq C_2|g(n)|\}$

If  $f \in \Theta(g)$ ,  $f$  grows (exactly) same order of  $g$ .

$f \in \Theta(g)$ , iff  $f \in O(g) \wedge f \in \Omega(g)$ .

$$|f(n)| \leq |O(g(n))| \quad |f(n)| \geq |\Omega(g(n))| \quad |\Theta(g(n))| \leq |f(n)| \leq |\Theta(g(n))|$$

Example 1, 2, 3, 4

**Theorem 4** Let  $f(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$  with  $a_n \neq 0$ . Then

$$f(x) \in \Theta(x^n).$$

## ***Computational complexity***

*time complexity*

*space complexity*

<i>Worst-case analysis</i>	$O(g(n))$
<i>Best-case Analysis</i>	$\Omega(g(n))$
<i>Average-case analysis</i>	$\Theta(g(n))$

$\Theta(1)$	<b><i>constant complexity</i></b>
$\Theta(\log n)$	<b><i>logarithmic complexity</i></b>
$\Theta(n)$	<b><i>linear complexity</i></b>
$\Theta(n^k)$	<b><i>polynomial complexity</i></b>
$\Theta(k^n)$	<b><i>exponential complexity</i></b>
$\Theta(n!)$	<b><i>factorial complexity</i></b>



**(totally) solvable**

**tractable**

*polynomial*

**intractable**

*exponential*

**class P**

*polynomial, tractable*

**class NP ( $P \subseteq NP$ )**

*Nondeterministic polynomial*

*exponential(upper bound)*

*lower bound?*

*tractable or intractable*

**NP-complete problems**

*subset of class NP*

*If any of these problems can be solved in polynomial time,*

*all the problems in class in **NP** also in class **P** ( $P = NP$ ).*