

## § 10 CONSTRUCTING AND EVALUATING PRESELECTIVE POLICIES

Systematic construction similar to ES-policies along a conflict settling tree

root = G

nodes = AND-OR networks arising from choices of waiting jobs  
on some forbidden sets

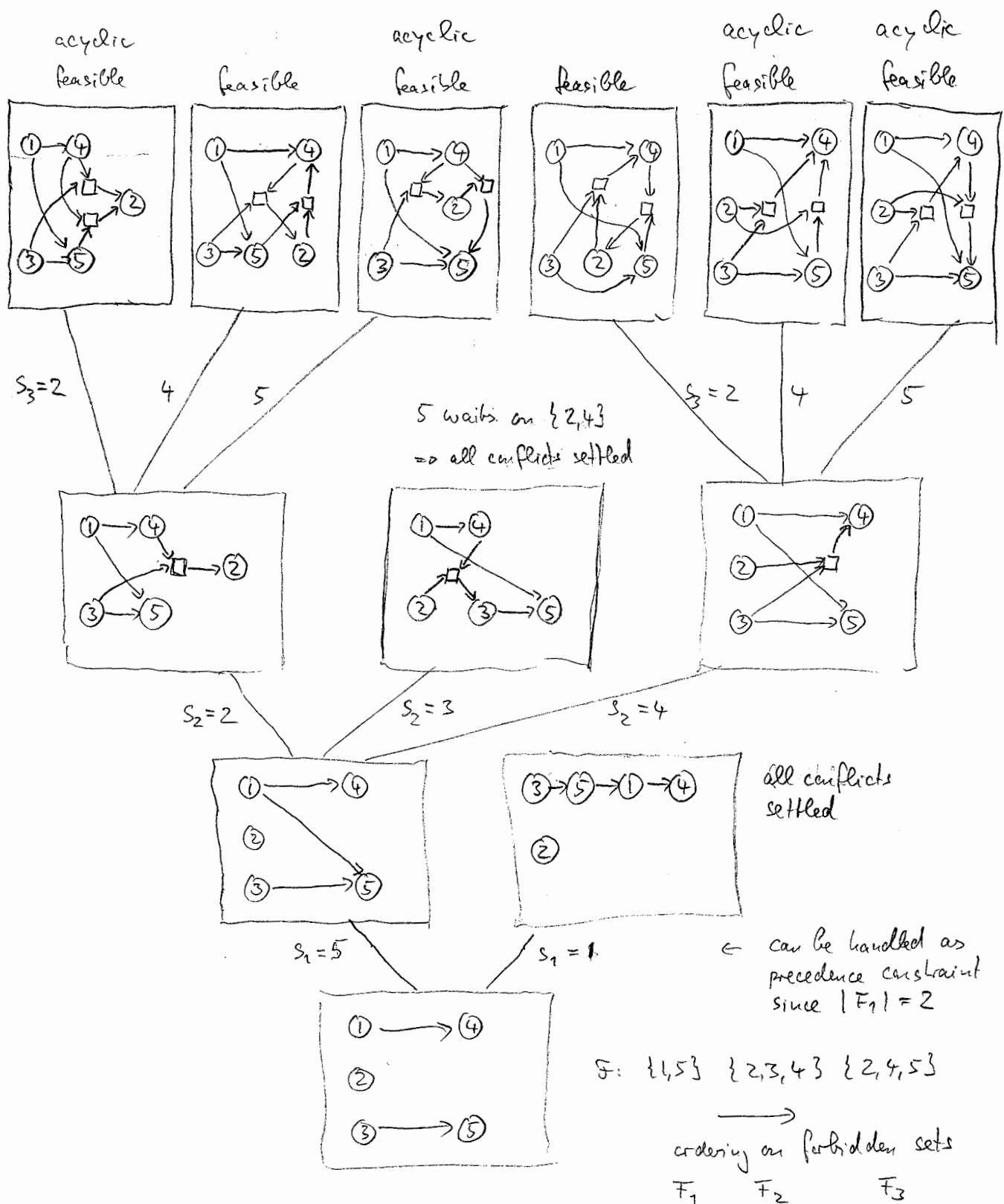
children of a node D = all AND-OR networks obtained from  
D by choosing a waiting job from  
one yet unsettled forbidden set

need to check this

may use suitable  
ordering of forbidden  
sets

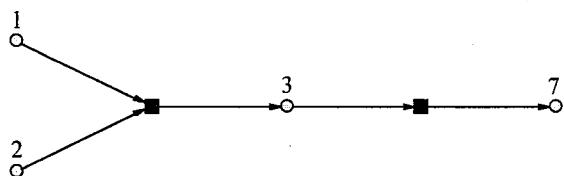
## 9.1 EXAMPLE continued

conflict settling tree



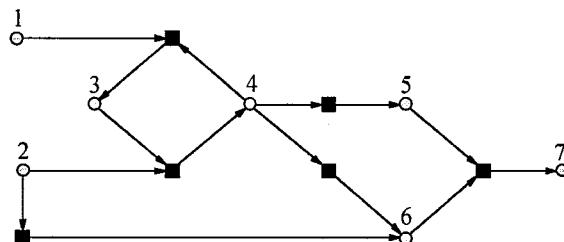
Checking if a forbidden set  $F$  is already settled by the partial selection  $(s_1, \dots, s_r)$  corresponds to finding forced waiting conditions of the form  $(F \setminus \{j\}, j)$

### Finding forced waiting conditions



easy to answer here

Is  $(\{1,2\}, 7)$  always implied by the given system  $\mathcal{W}$ ?



not so easy  
in this case

Def:  $j$  forced to wait for  $u \Leftrightarrow$  all linear realizers have some  $u \sqsubset$  before  $j$

### An algorithm for finding forced waiting conditions

- ◆ **Input:** Jobs  $V$ , feasible waiting conditions  $\mathcal{W}$ , set  $U \subseteq V$
- ◆ **Output:** A list  $L$  of jobs

List  $L := []$

**while** (there is a job  $i \in V \setminus U$  that is not a waiting job in  $\mathcal{W}$ )

**begin**

    insert  $i$  at the end of  $L$

**if** (some waiting condition  $(X, j)$  becomes satisfied)  
        delete  $(X, j)$  from  $\mathcal{W}$

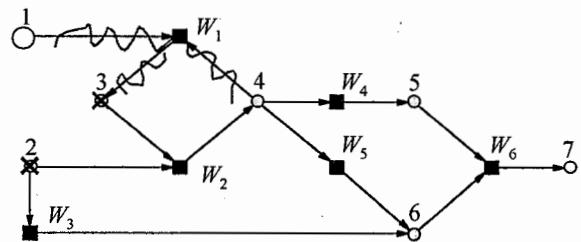
**end**

**return**  $L$

delete all such  
waiting conditions

First example

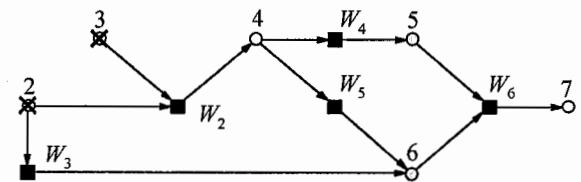
$$U = \{2, 3\}$$



$$L = [1, \quad ]$$

First example

$$U = \{2, 3\}$$

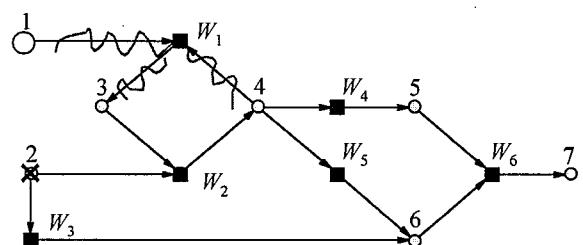


$$L = [1, \quad ] \Rightarrow \text{termination}$$

Claim: Every job not in  $L \cup U$  waits for  $U = \{2, 3\}$

### Second example

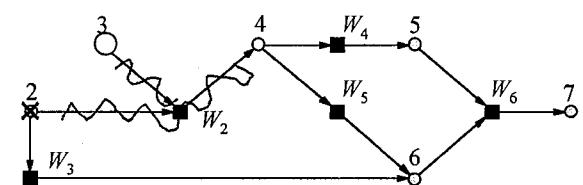
$$U = \{2\}$$



$$L = [1, ]$$

### Second example

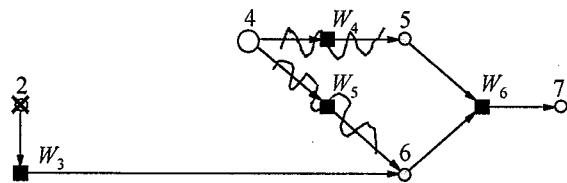
$$U = \{2\}$$



$$L = [1, 3, ]$$

### Second example

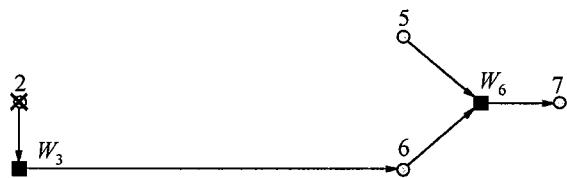
$$U = \{2\}$$



$$L = [ 1, 3, 4, ]$$

### Second example

$$U = \{2\}$$



$$L = [ 1, 3, 4, ] \Rightarrow \text{termination with } L = [ 1, 3, 4, 5, 7 ]$$

Claim: 6 waits for  $U = \{2\}$

## Correctness of the algorithm

$(U, j)$  is a forced waiting condition  $\Leftrightarrow j \notin L$  (and  $\notin U$ )

10.2 THEOREM

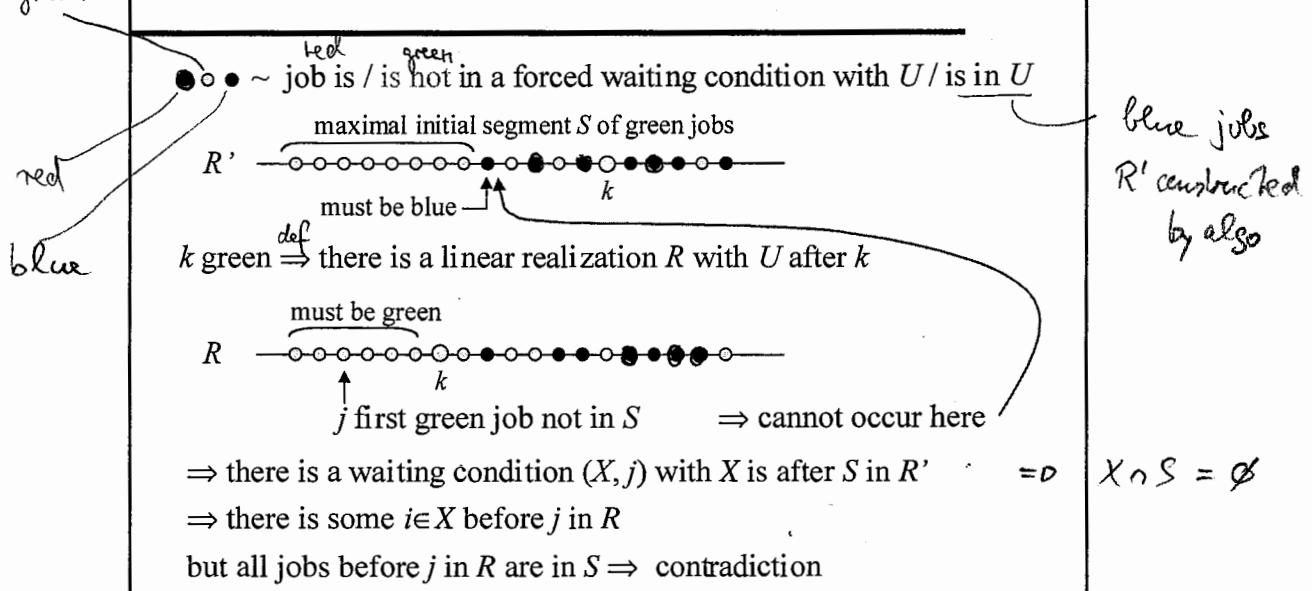
Key lemma:

There is a linear realization of  $\mathcal{W}$  starting with all jobs  $j$  for which  $(U, j)$  is not a forced waiting condition

10.3 LEMMA

$\Rightarrow$  only  $U$  and forced waiting jobs are not in  $L$  at termination of the algorithm

## Proof of key lemma (by contradiction)



Summary: Can detect forced waiting conditions in linear time

## Computing earliest start times for a preselective policy

Input: Preselective policy  $\Pi$ , processing time vector  $x$

Output: Vector  $\Pi[x]$  of start times

= Earliest start w.r.t. system of waiting conditions given by selection  $s$  defining  $\Pi$  and graph  $G$  of precedence constraints

translate this to algorithm on the AND/OR-network representing  $\Pi$  and  $G$ .

↓ § 9

Solve a system of min-max inequalities and compute the unique componentwise minimal solution

Special case needed here

have positive arc weights

$d_{jw} = x_j$  for arcs  $(j) \rightarrow [w]$

may assume arc weight

$d_{wj} = 0$  for arcs  $[w] \rightarrow (j)$

since only one outgoing arc from every OR-node

general case:

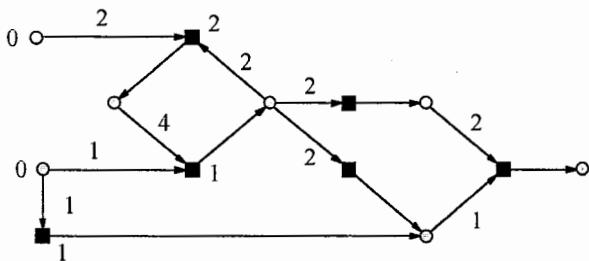
arbitrary arc weights

$d_{jw}$ ,  $d_{wj}$

(also negative)

Solve the special case by a Dijkstra-like algorithm:

### A Dijkstra-like algorithm for positive arc weights



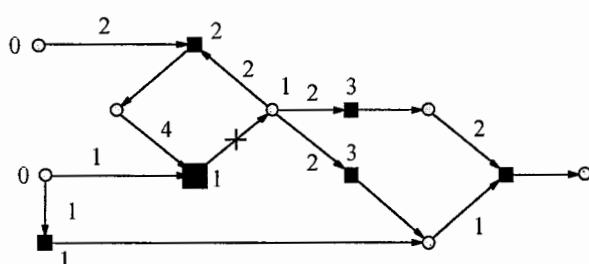
Assume w.l.o.g that  $d_{wj} = 0$  (only one outgoing arc per OR node)  
and that waiting conditions are feasible

set  $S_j := 0$  if there is no  $(w, j) \in A$

for unmarked OR nodes  $w \in \text{out}(j)$ , set  $S_w = \min\{S_j + d_{jw} \mid (j, w) \in A\}$   
set  $S_w = \infty$  otherwise

### 10.4 ALGORITHM

### A Dijkstra-like algorithm for positive arc weights



choose unmarked OR-node  $w = (X, j)$  with minimum  $S_w$  and mark  $w$   
reduce indegree of  $j$  by 1

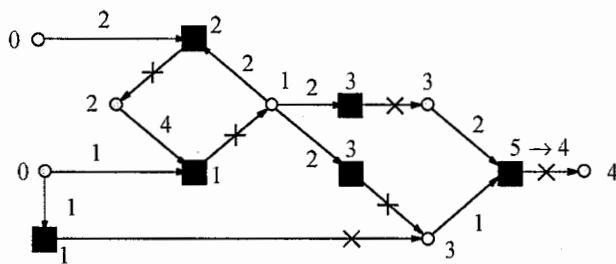
if indegree( $j$ ) = 0 then

set  $S_j := \max\{S_w \mid (w, j) \in A\}$

for unmarked OR nodes  $w \in \text{out}(j)$ , set  $S_w = \min\{S_j + d_{jw} \mid (j, w) \in A\}$

big  $\blacksquare$   $\triangleq$  marked  
 $\rightarrow$   $\triangleq$   
 repeat  
 reducing indegree

### A Dijkstra-like algorithm for positive arc weights



$5 \rightarrow 4 \quad \text{?}$   
change of  
distance at  
OR-node

choose OR-node  $w = (X, j)$  with minimum  $S_w$  and mark  $w$   
reduce indegree of  $j$  by 1  
if  $\text{indegree}(j) = 0$  then  
  set  $S_j := \max\{S_w \mid (w, j) \in A\}$   
  for unmarked OR nodes  $w \in \text{out}(j)$ , set  $S_w = \min\{S_j + d_{jw} \mid (j, w) \in A\}$

repeat

10.5 THEOREM : Algorithm 10.4 computes the unique minimal feasible solution  $\geq 0$  of the min-max system given by the AND/OR graph  $D = (V \cup W, A)$  in  $O(|V| + |W| \cdot \log |W| + |A|)$  time

Proof: (1) Correctness:

let  $S$  be the vector of start times constructed by the algorithm.

let  $S^*$  be the ES-vector (see Lemma 9.6) of  $(V, W)$

Assume that  $S \neq S^*$ .

Then choose node  $v$  with  $S_v > S_v^*$  and  $S_v^*$  minimum.

Case 1:  $v$  is an AND-node

$\Rightarrow \exists$  OR-node  $w = (X, v)$  with  $S_w = S_v + \underbrace{d_{vw}}_0$

$\boxed{w} \longrightarrow \textcircled{v} \Rightarrow S_w = S_v > S_v^* \geq S_w^*$

$\Rightarrow$  (choice of  $v$ )  $S_v^* = S_w^*$  and  $S_w > S_w^*$

$\Rightarrow$  reduced to the case that  $v$  is an OR-node

Case 2:  $v$  is an OR-node

$\Rightarrow \exists$  AND-node  $i$  with  $S_v^* = S_i^* + \underbrace{d_{iv}}_{>0} \quad (*)$



Claim:  $S_i > S_i^*$

Suppose not, i.e.  $S_i = S_i^*$

then, when the algorithm assigns to  $i$  the value  $S_i$ ,

$S_v$  is set to  $\min_{(j,v)} \{S_j + d_{iv}\} \leq S_i^* + d_{iv} = S_v^*$

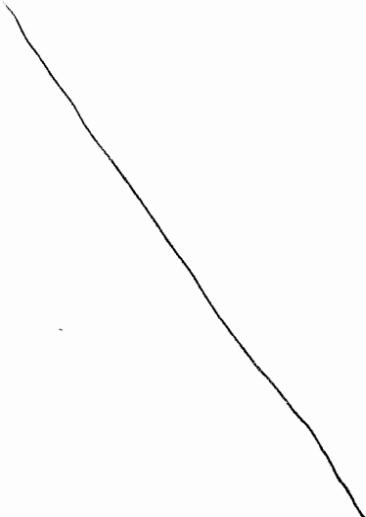
(if unmarked), or  $S_v \leq S_i + d_{iv} = S_i^* + d_{iv} = S_v^*$

(if already marked)

$\Rightarrow$  contradiction to  $S_v > S_v^*$

So  $S_i > S_i^*$  and  $S_i^* < S_v^*$  because of  $(*)$

$\Rightarrow$  contradiction to the choice of  $v$   $\square$



(2) Runtime: Exercise  $\square$

For a min-max system,  $(\infty, \infty, \dots, \infty)$  is a solution.

We call a solution  $S$  feasible, if every  $s_j < \infty$ , and the min-max system feasible if there is a feasible solution.

10.6 LEMMA: A min-max system with  $x_{jw} > 0$  and  $x_{wj} \geq 0$  has a feasible solution  $S$  iff the weighting conditions are feasible (feasibility of min-max system = structural feasibility).

Proof: " $\Rightarrow$ " all  $x_{jw} > 0 \Rightarrow$  for every  $w = (X, j)$  there is an  $i \in X$  with  $s_i < s_w \leq s_j$ .

(otherwise  $S$  is not feasible)

The arcs  $(i, j)$  define a realization of  $W$ .

If not, they contain a cycle, and we would have  $s_\ell < s_k$  for every arc  $(\ell, k)$  on the cycle, which cannot be the case.

$\Leftarrow$   $W$  feasible  $\Rightarrow \exists$  realization  $R$ .  $E_R$  defines a feasible solution of the min-max system  $\square$

Remark: For  $d_{jw} \geq 0$ ,  $d_{wj} \geq 0$ , structural feasibility is not equivalent to feasibility of the min-max system

Still possible to decide feasibility in polynomial time (Exercise)

The general case: arbitrary  $d_{jw}$  and  $d_{wj}$

### Some first observations

$$\left. \begin{array}{l} j \in V \text{ AND node: } S_j \geq \max_{(w,j) \in A} [S_w + d_{wj}] \\ w \in \mathcal{W} \text{ OR node: } S_w \geq \min_{(j,w) \in A} [S_j + d_{jw}] \end{array} \right\} \text{min-max system}$$

$(\infty, \infty, \dots, \infty)$   
is a solution

$(S_1, \dots, S_n)$  and  $(T_1, \dots, T_n)$  solutions  
 $\Rightarrow (\min\{S_1, T_1\}, \dots, \min\{S_n, T_n\})$  solution

- ◆ There is a unique componentwise minimal solution  $S \geq 0$
- ◆ Problem is feasible iff every  $S_j < \infty$
- ◆ There is a unique maximal feasible subset of jobs

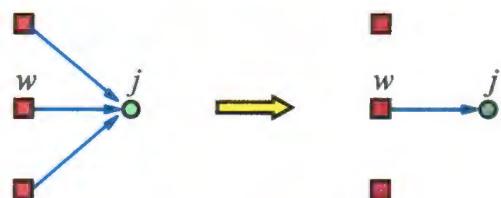
## Certificate for SOLVABILITY $\in \text{NP}$

Any feasible schedule  $S = (S_1, \dots, S_n)$  is a certificate for SOLVABILITY  $\in \text{NP}$

## Certificate for SOLVABILITY $\in \text{coNP}$

Let  $S = (S_1, \dots, S_n)$  be the unique minimal solution (maybe with  $\infty$ )

For every AND node  $j$ ,  
one can delete  
all but one incoming arcs  
without changing  $S_j$



needs  
small  
proofs

Then every cycle has non-negative length

Relaxing in every AND-node  
 $\Rightarrow$  relaxed problem with only min inequalities ( $\sim$  OR nodes)  
 $\Rightarrow$  can check  $S_j > K$  by shortest path algorithms in polynomial time  
 $\Rightarrow$  relaxed problem is certificate for SOLVABILITY  $\in \text{coNP}$

Proof: relaxing

delete all possible st.  $S^*$  is not changed

Suppose  
 $\exists j$  with  
 indegree  $> 1$



$\Rightarrow$  if let  $S^1$  be the best schedule  
 after deleting  $(w_1, j)$

$$S^1 \dots (w_2, j)$$

$$\Rightarrow S_j^* > S_j^1 \quad S_j^* > S_j^2 \quad \text{and w.l.o.g. } \underline{S_j^1 \leq S_j^2}$$

$$\text{Set } S_i := \min \{ S_i^1 + S_i^2 - S_j^1, S_i^2 \}$$

$$\Rightarrow S \leq S_2 \quad , \quad S_j = S_j^2$$

if nodes (AND, OR)  
 show that  $S$  is a schedule

$$\text{Look at } \boxed{w} \xrightarrow{\text{O}} \boxed{j} \quad w \neq w_2$$

$$\Rightarrow S_j^2 \geq S_w \quad \Rightarrow \quad S_j = S_j^2 \geq S_w$$

$$\boxed{w_2} \xrightarrow{\text{O}} j \quad \Rightarrow \quad S_j = \underbrace{S_j^1 + S_j^2 - S_j^1}_{\geq S_{w_2}^1 \geq 0} \geq S_{w_2}$$

$$\Rightarrow S_j \geq \max_{(w,j)} S_w$$

For other AND nodes ( $\neq_j$ )

$$\boxed{w} \xrightarrow{\text{O}} \boxed{k} \quad S_k^1 \geq S_w^1$$

$$S_k^2 \geq S_w^2$$

$$\left. \begin{array}{l} \\ \end{array} \right\} \Rightarrow S_k^1 + S_k^2 - \overbrace{S_j^1}^{\text{keinbest}} \geq S_w^1 + S_w^2 - S_j^1$$

Similar for OR nodes

$$\Rightarrow S = \min \{ S^1 + S^2 - S_j^1, S^2 \} \text{ is a schedule}$$

$$\text{contradiction to } S_j = S_j^2 < S_j^*$$

$\uparrow$  best schedule by construction

## Complexity of checking solvability

A schedule and a tightened subproblem are polynomially checkable certificates for membership in NP and coNP



SOLVABILITY  $\in \text{NP} \cap \text{coNP}$

T6.7 THEOREM

No polynomial algorithm known

Not known to be NP-complete / coNP complete

Same complexity status as

LINEAR PROGRAMMING



LP by Ellipsoid  
Method

PRIMES



solved 2002,  $\in \text{P}$   
AKS primality test  
Agrawal - Kayal - Saxena

Exercises:

10.1 Show that Algorithm 10.4. can be implemented to run in  
 $O(|V| + |W| \cdot \log |W| + |A|)$  time

10.2\* Derive a polynomial-time algorithm for finding the unique minimal (feasible) solution  $\geq 0$  of a min-max system with non-negative arc weights

Hint: Try to relate feasibility of the min-max system to structural feasibility in the sense of Lemma 10.6.  
What changes?

10.3 Derive a pseudo-polynomial-time algorithm for finding the unique minimal (feasible) solution  $\geq 0$  of a min-max system with arbitrary arc weights