

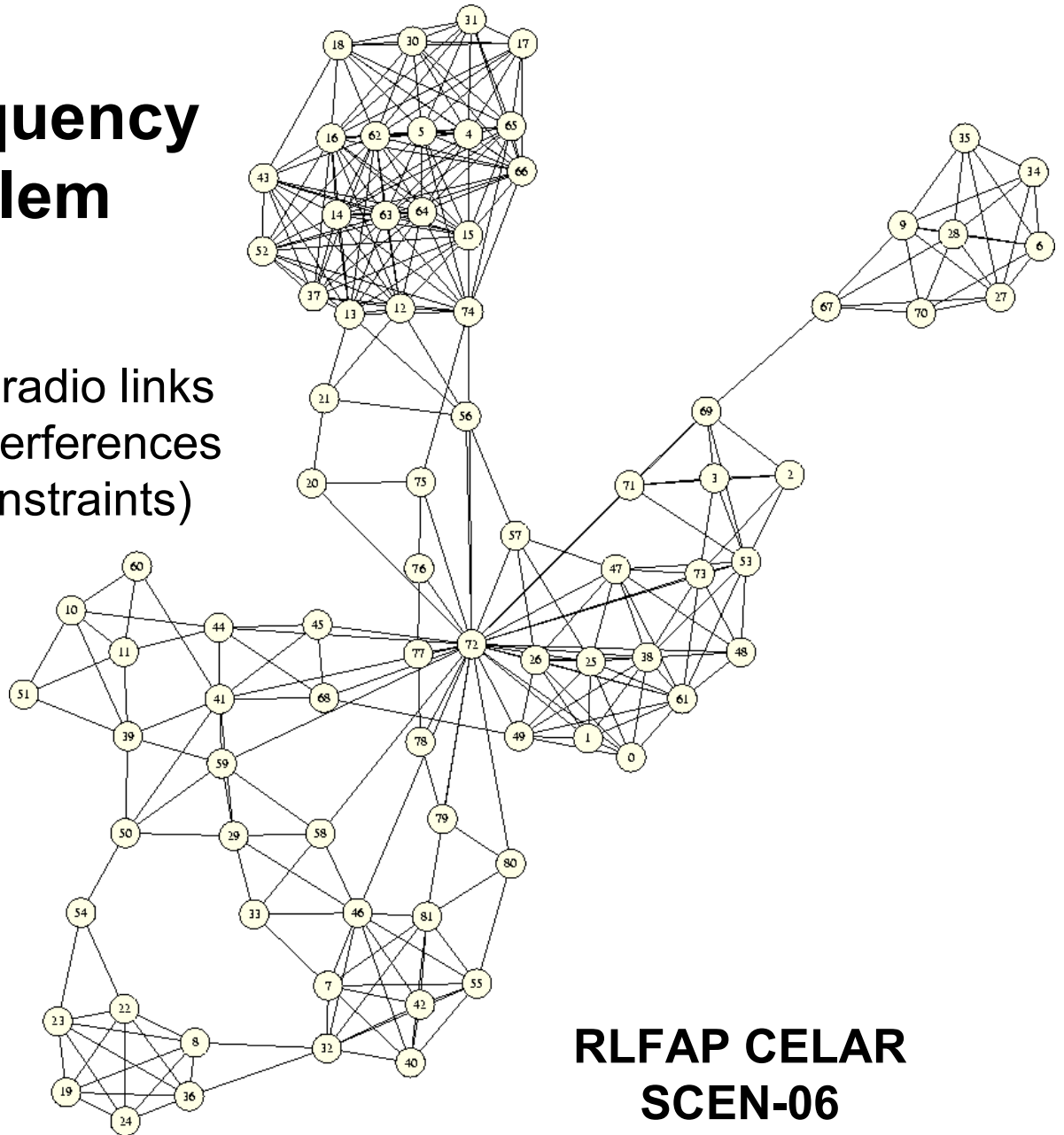


# **Exploiting Tree Decomposition and Soft Local Consistency in Weighted CSP**

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Toulouse, France**

# Radio Link Frequency Allocation Problem

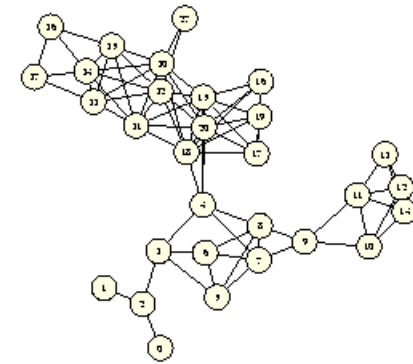
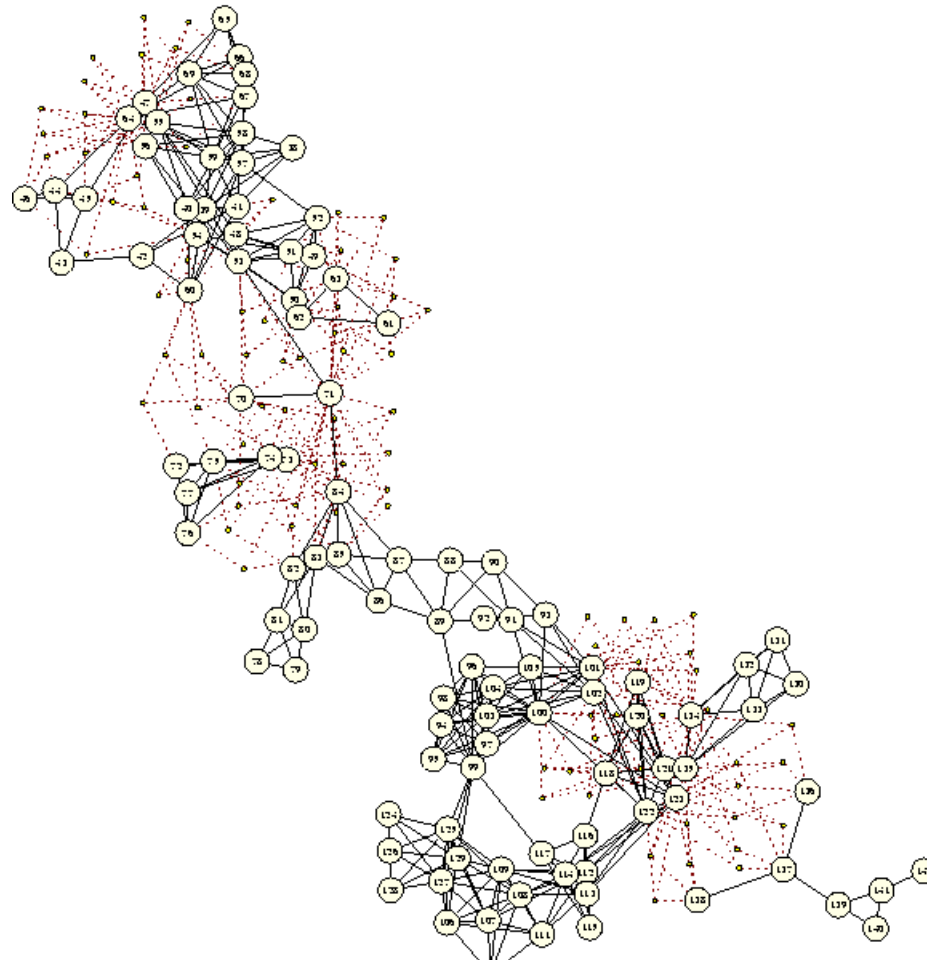
Allocate frequencies to radio links such that the sum of interferences (soft binary distance constraints) is minimized



**RLFAP CELAR  
SCEN-06**

# Earth Observation Satellite Management Problem

Select images among a set of candidate images such that physical constraints (hard binary and ternary constraints) are satisfied and the sum of weights associated to selected images (soft unary constraints) is maximized



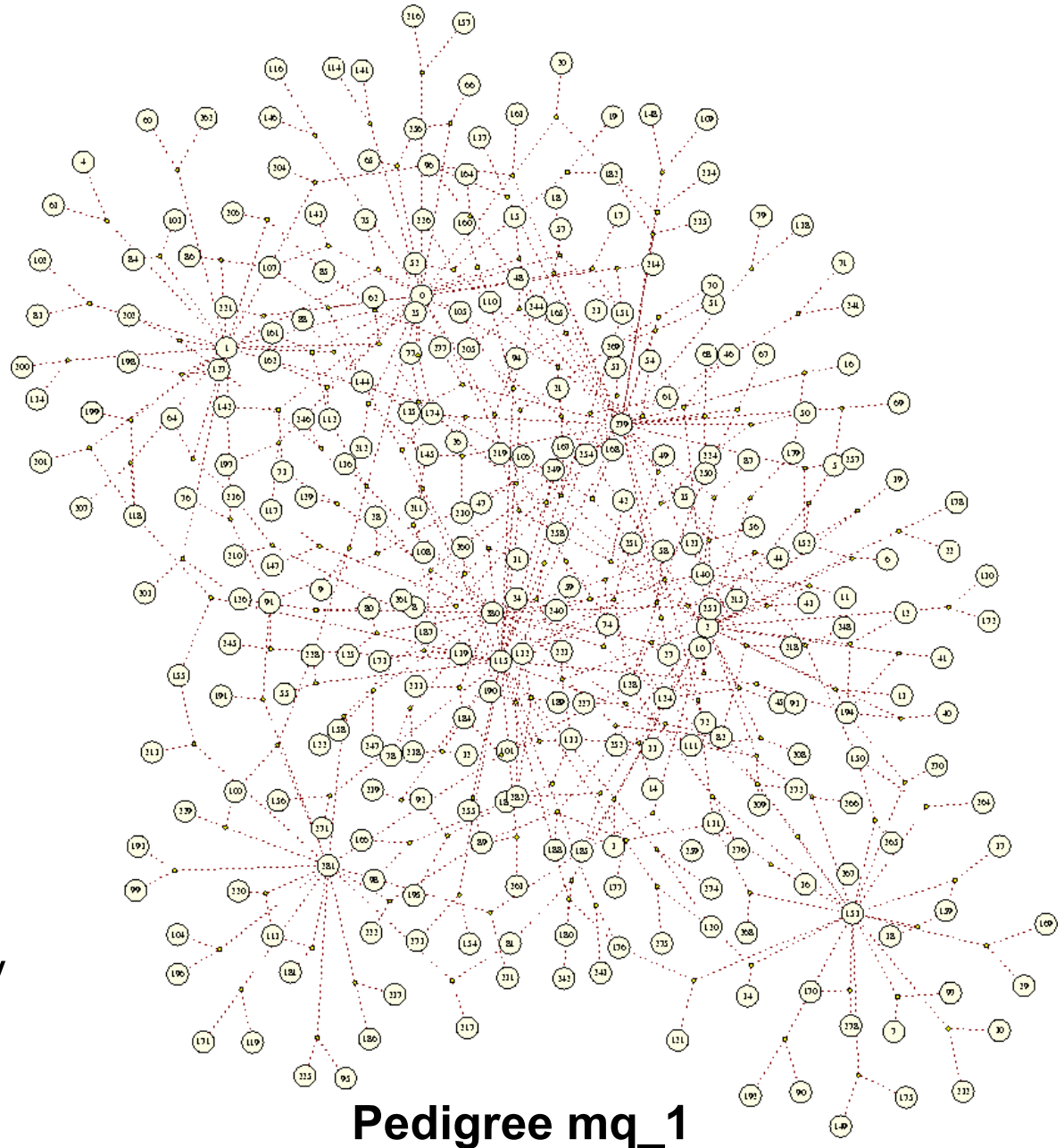
**SPOT5 #503**



# Mendelian error detection in complex pedigree

Find a complete genotype of maximum a posteriori probability (MPE)

Conditional probability tables (soft unary, binary and ternary constraints)





## Motivation of this work

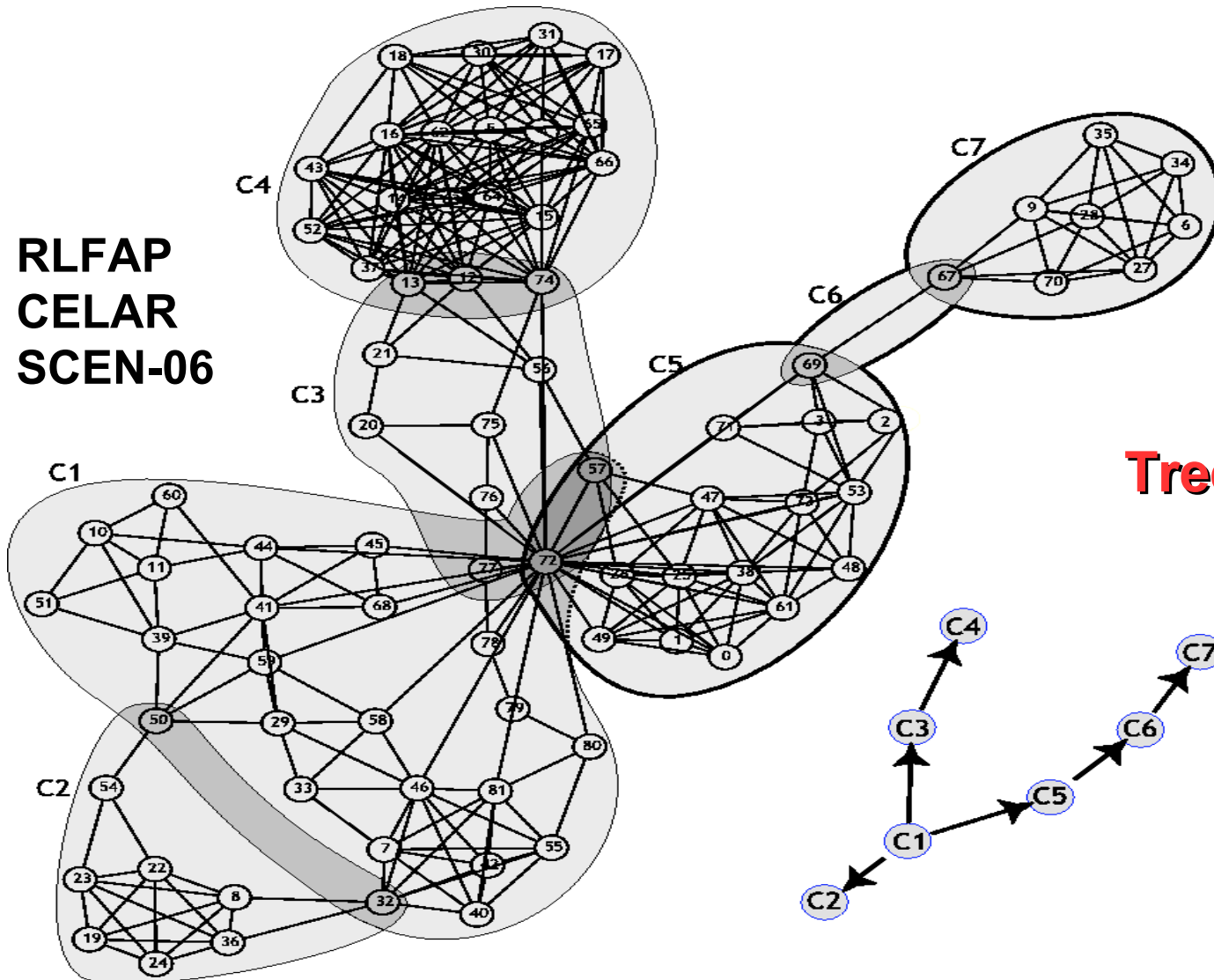
**Exploitation of the structure**  
present in many real problems  
**temporal, spatial, causal ...**

# Framework: **weighted binary CSP**

- $(X, D, W)$ 
  - $X = \{x_1, \dots, x_n\}$  **n variables**
  - $D = \{D_1, \dots, D_n\}$  **n finite domains** of maximum size **d**
  - $W = \{W_1, \dots, W_e\}$  **e cost functions**
    - $W_{ij}, W_i, W_\emptyset$  with **scopes**  $\{x_i, x_j\}, \{x_i\}, \emptyset$
    - $W_{ij} : D_i \times D_j \rightarrow [0, k]$
    - **k** is associated with completely **forbidden assignments**
- Find a **complete assignment** minimizing
$$W_\emptyset + \sum_i W_i(a_i) + \sum_{ij} W_{ij}(a_i, a_j)$$
- **NP-hard** problem

# Tree decomposition of a constraint graph

RLFAP  
CELAR  
SCEN-06



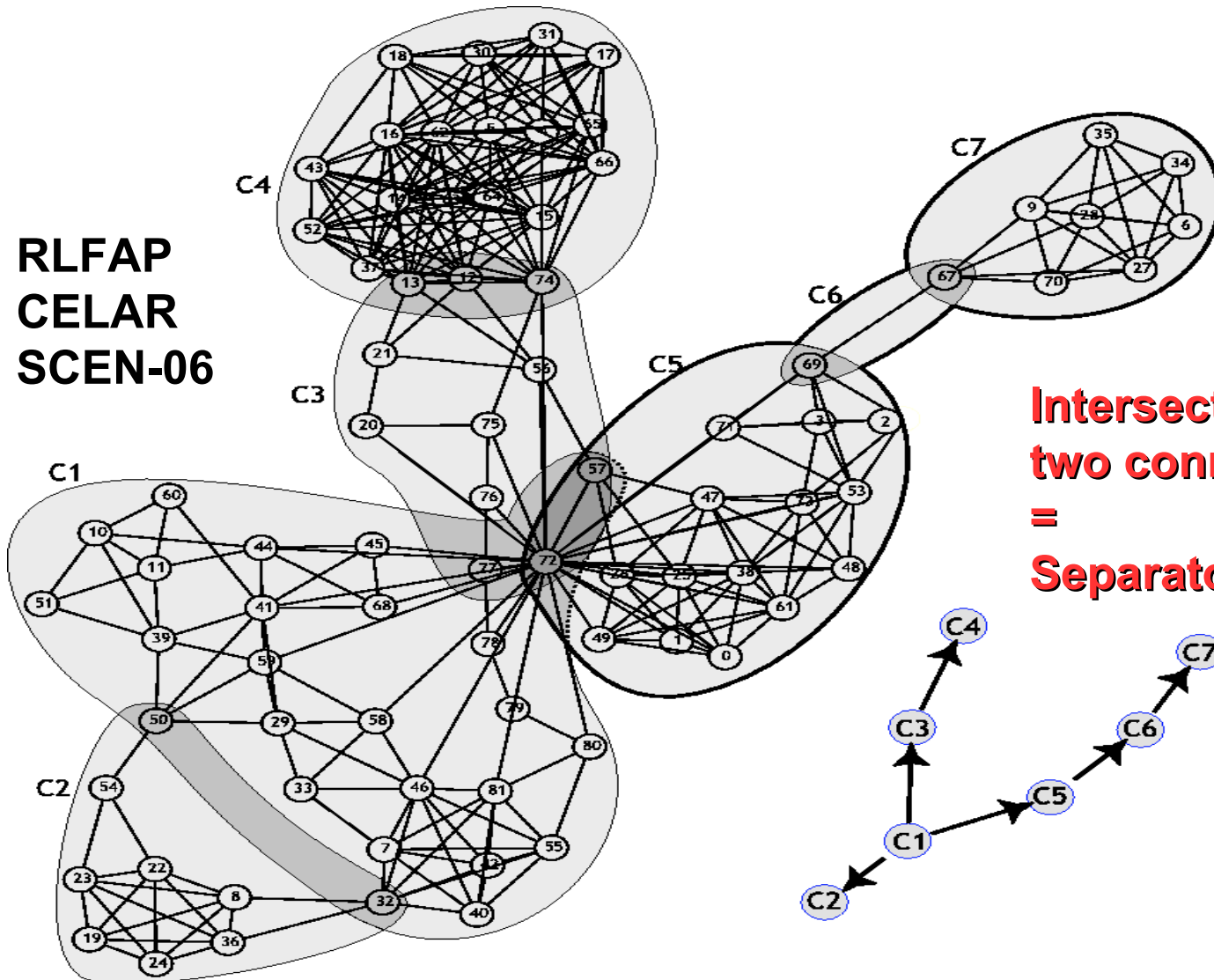
**Tree of clusters**





# Tree decomposition of a constraint graph

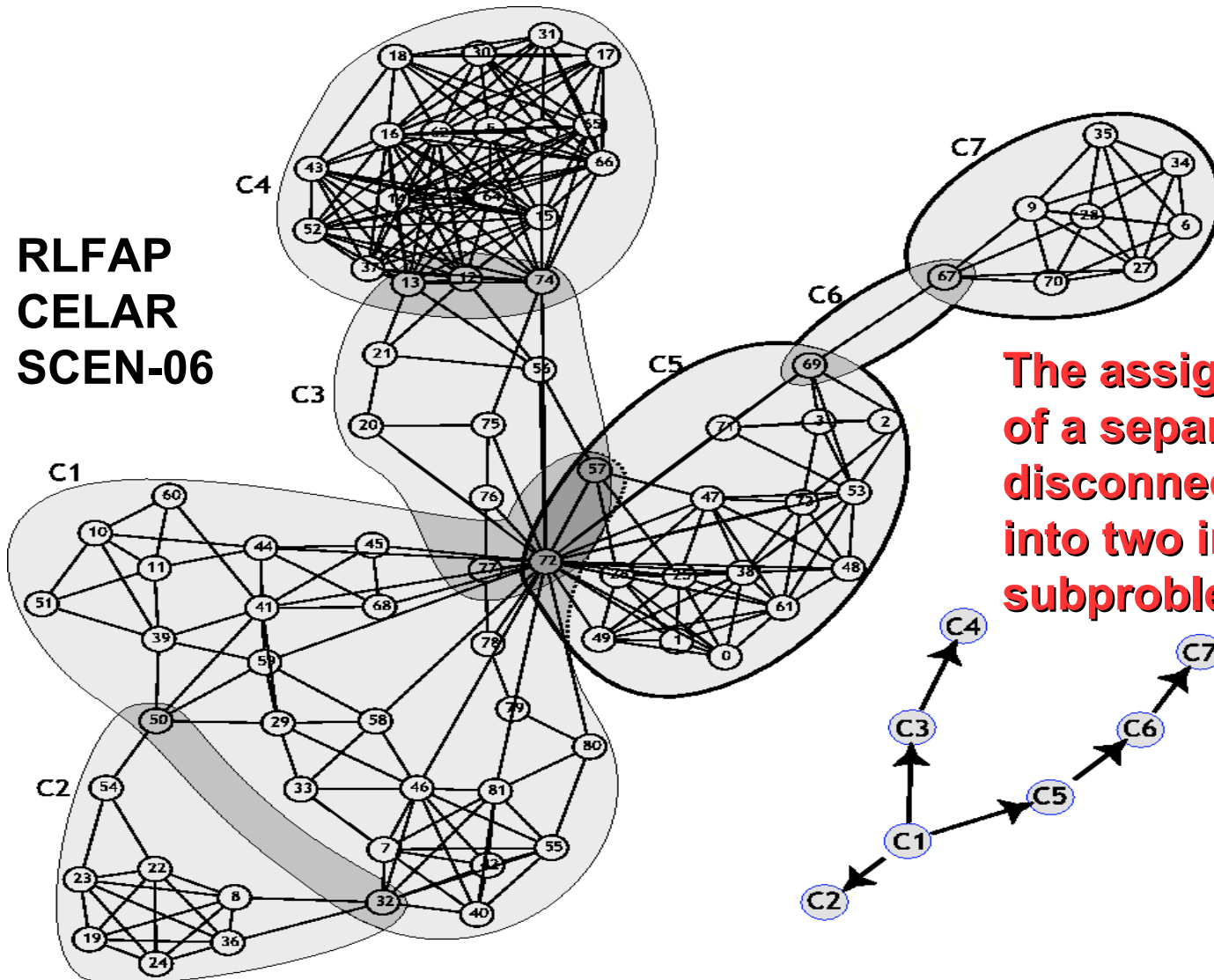
RLFAP  
CELAR  
SCEN-06



**Intersection between  
two connected clusters  
=  
Separator**

# Tree decomposition of a constraint graph

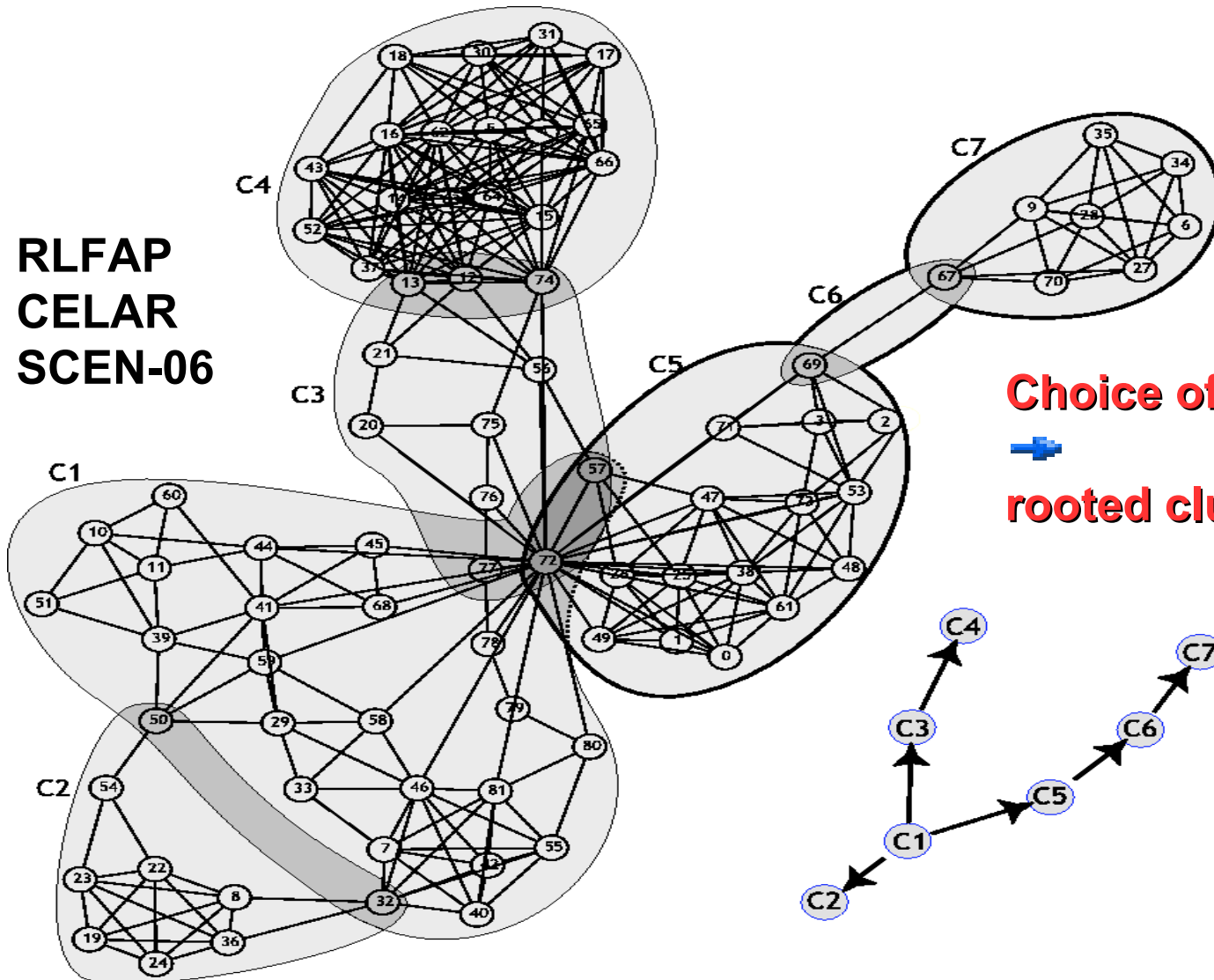
RLFAP  
CELAR  
SCEN-06



**The assignment of a separator disconnects the problem into two independent subproblems**

# Tree decomposition of a constraint graph

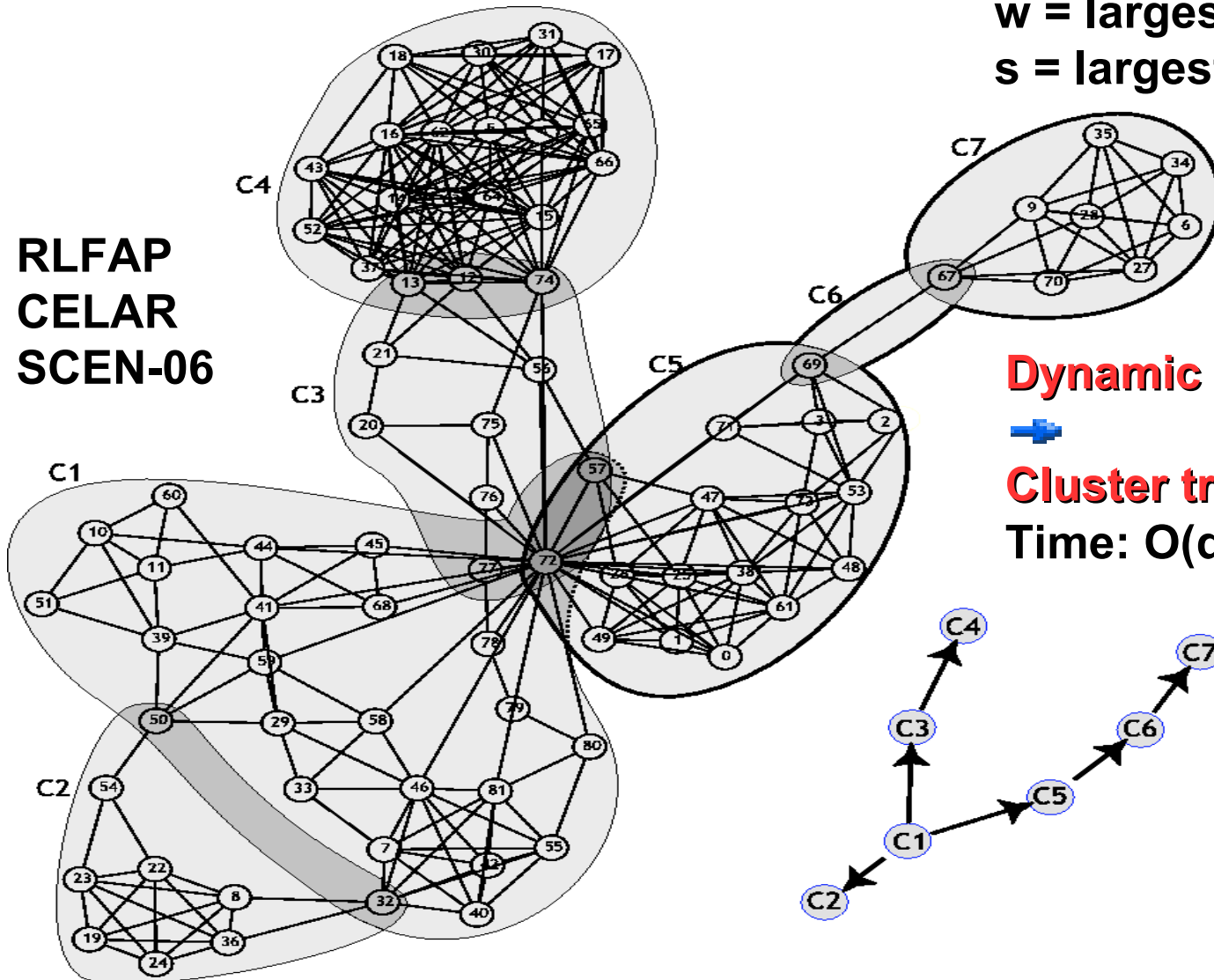
RLFAP  
CELAR  
SCEN-06



**Choice of a root cluster**  
➔  
**rooted cluster tree**

# Exploitation of the tree decomposition

RLFAP  
CELAR  
SCEN-06



$w$  = largest cluster size  
 $s$  = largest separator size

**Dynamic programming**

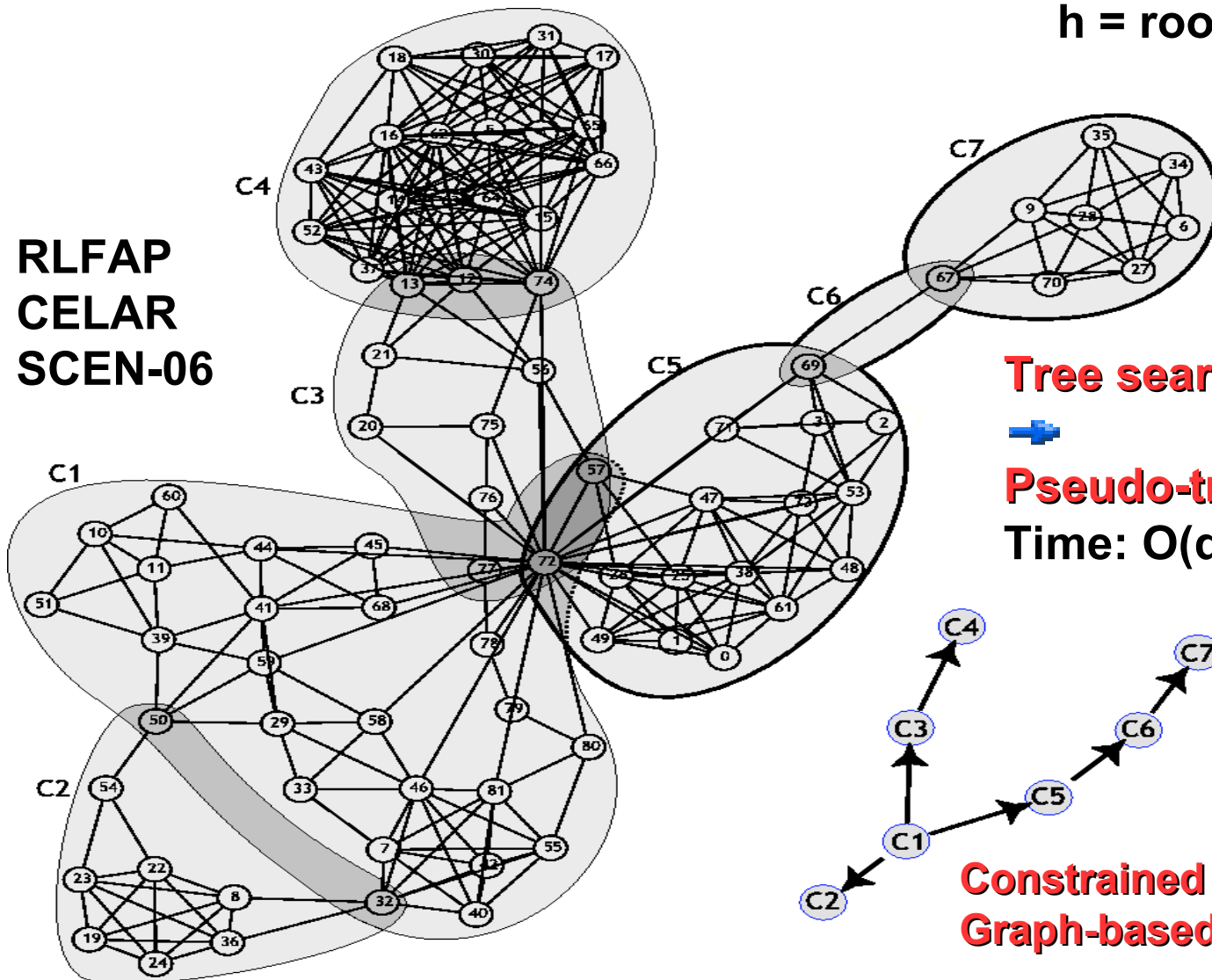


**Cluster tree elimination**

Time:  $O(d^w)$ , Space:  $O(d^s)$

# Exploitation of the tree decomposition

RLFAP  
CELAR  
SCEN-06



**Tree search**



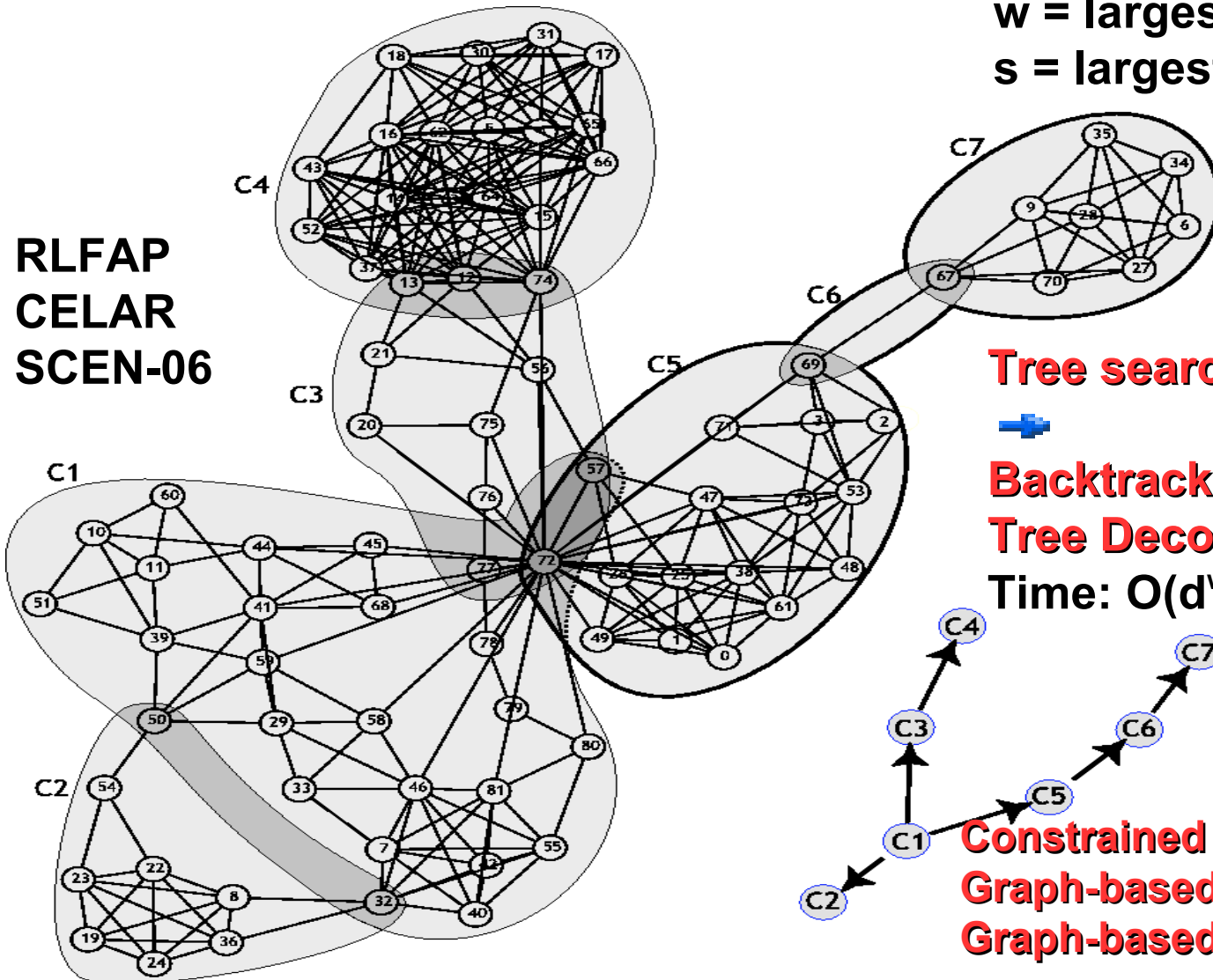
**Pseudo-tree search (PTS)**

Time:  $O(d^h)$ , Space:  $O(nd)$

**Constrained variable ordering**  
**Graph-based backjumping**

# Exploitation of the tree decomposition

RLFAP  
CELAR  
SCEN-06



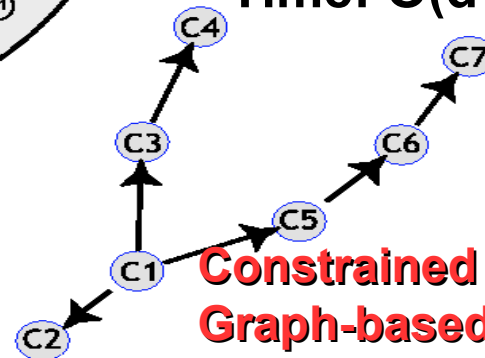
$w$  = largest cluster size  
 $s$  = largest separator size

**Tree search**



**Backtrack Bounded by  
Tree Decomposition (BTD)**

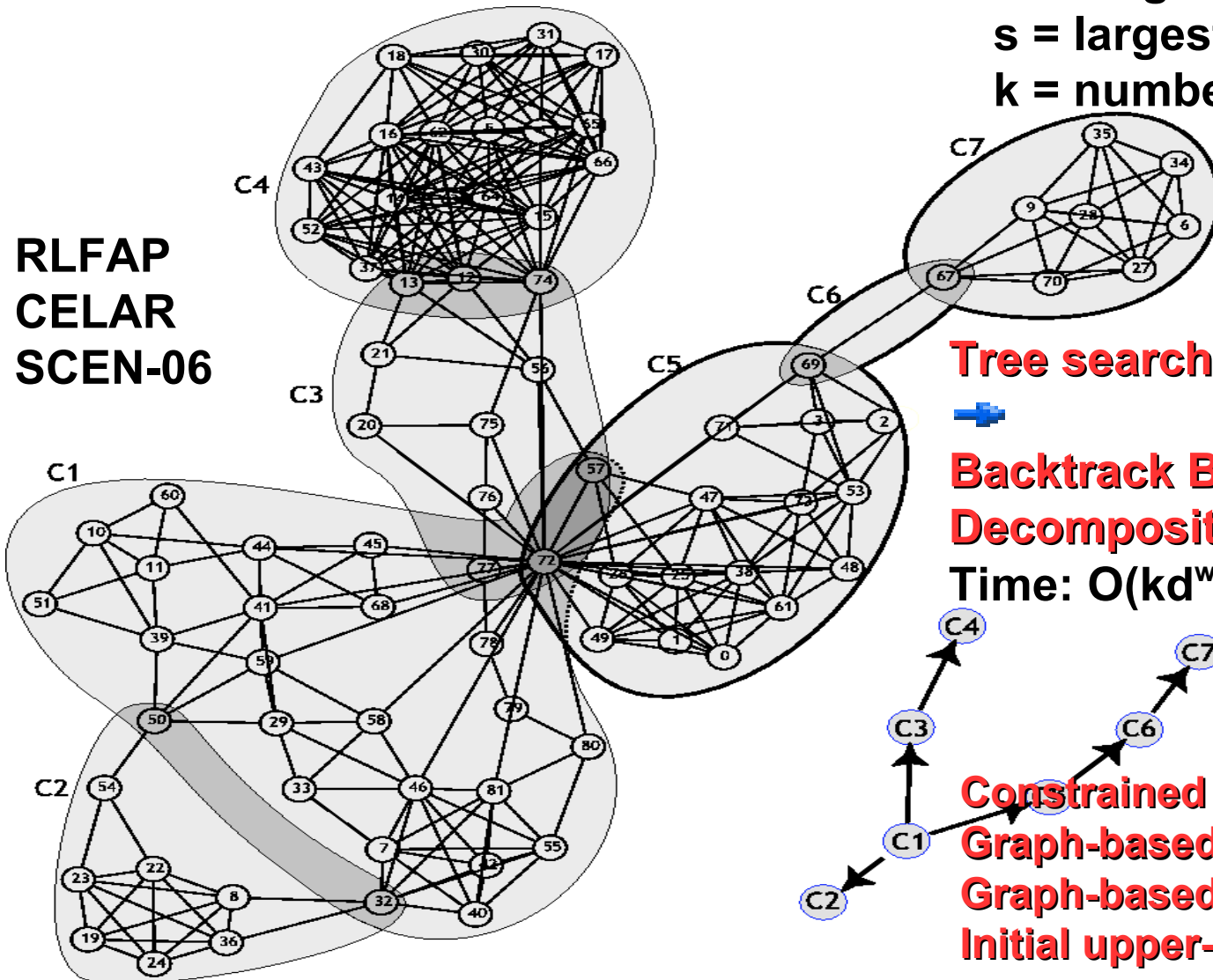
Time:  $O(d^w)$ , Space:  $O(d^s)$



**Constrained variable ordering  
Graph-based backjumping  
Graph-based learning**

# Exploitation of the tree decomposition

RLFAP  
CELAR  
SCEN-06



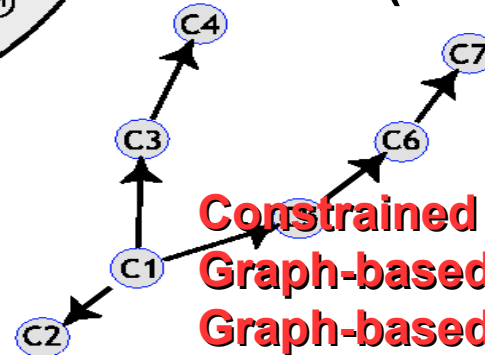
w = largest cluster size  
s = largest separator size  
k = number of valuations

**Tree search**



**Backtrack Bounded by Tree Decomposition (BTD+)**

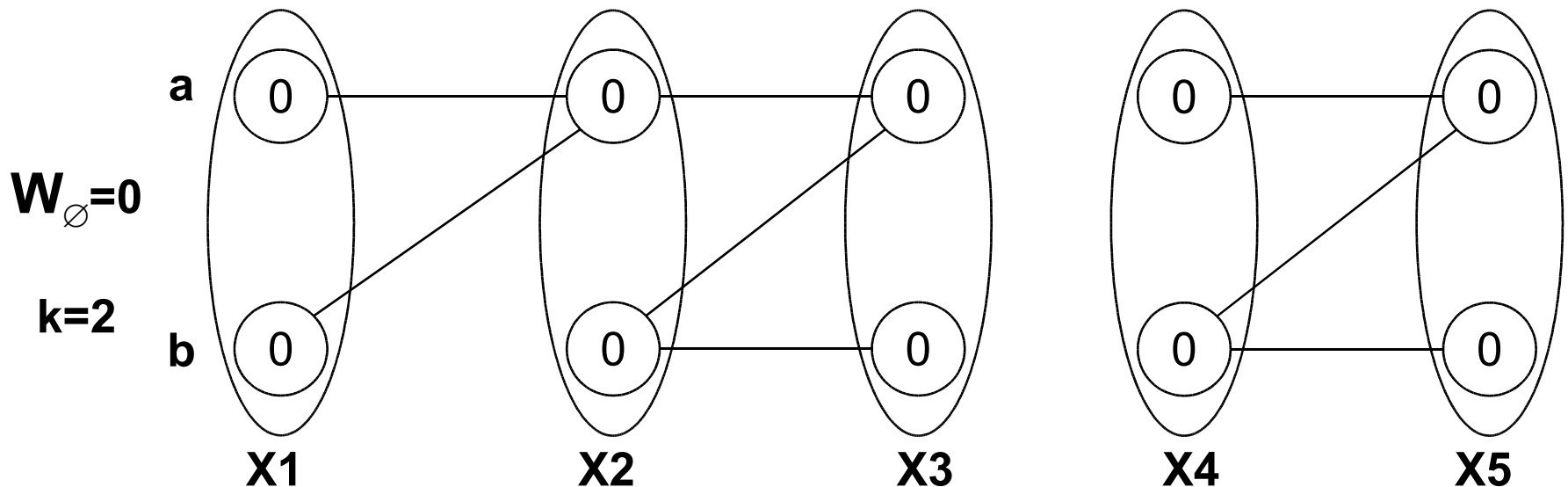
Time:  $O(kd^w)$ , Space:  $O(d^s)$



**Constrained variable ordering**  
**Graph-based backjumping**  
**Graph-based learning**  
**Initial upper-bounds**

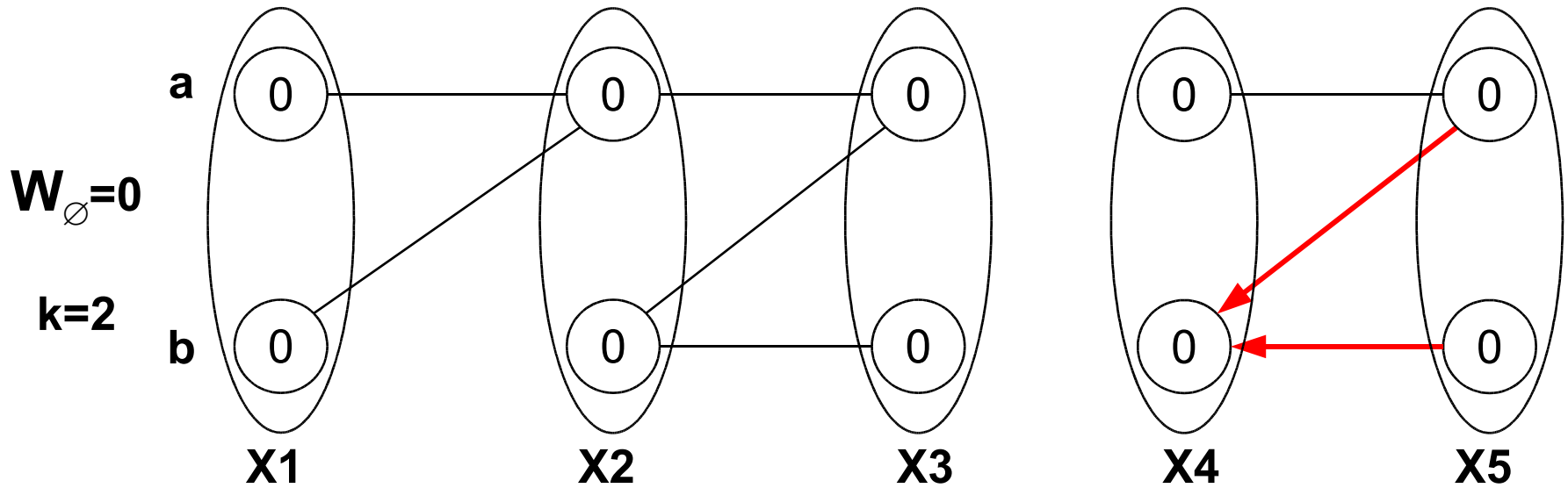
# Soft local consistency

- **Goal:** transforming a problem into an **equivalent problem** with a **more explicit lower bound**
- **Means:** enforcing a **soft local consistency property** by **moving costs** from **binary** constraints to **unary** constraints and to the **zero-arity** constraint  $W_{\emptyset}$  (problem lower bound)



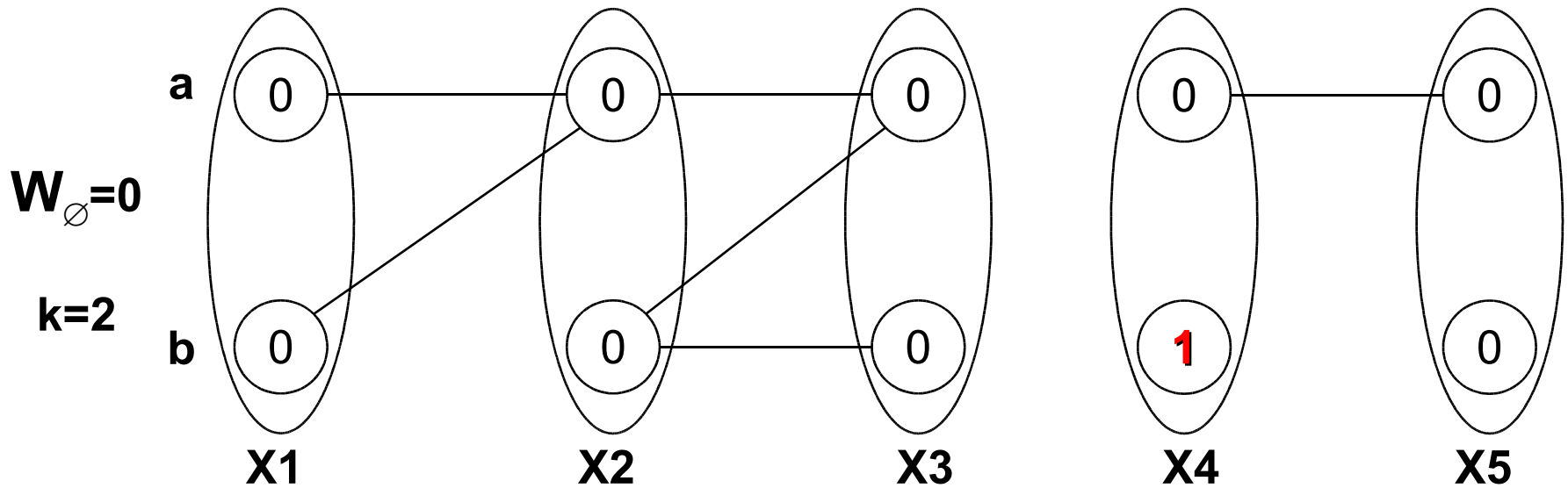


# Soft local consistency



**Cost projection  
from a binary to  
a unary constraint**

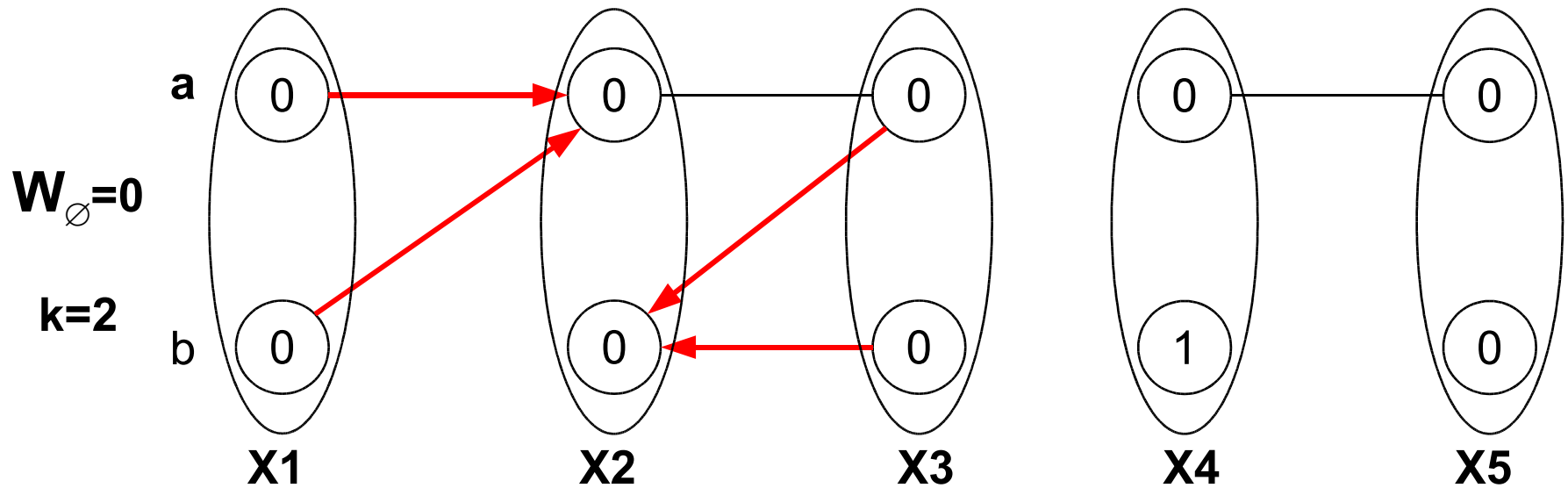
# Soft local consistency



No idempotency property

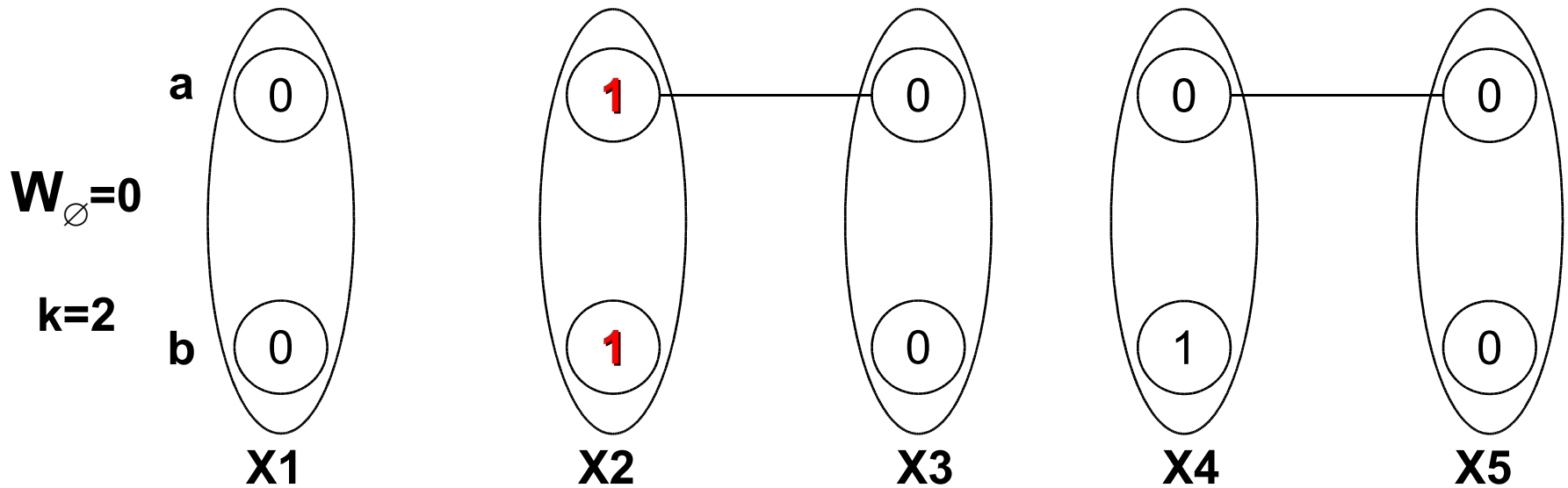
**Cost projection  
from a binary to  
a unary constraint**

# Soft local consistency



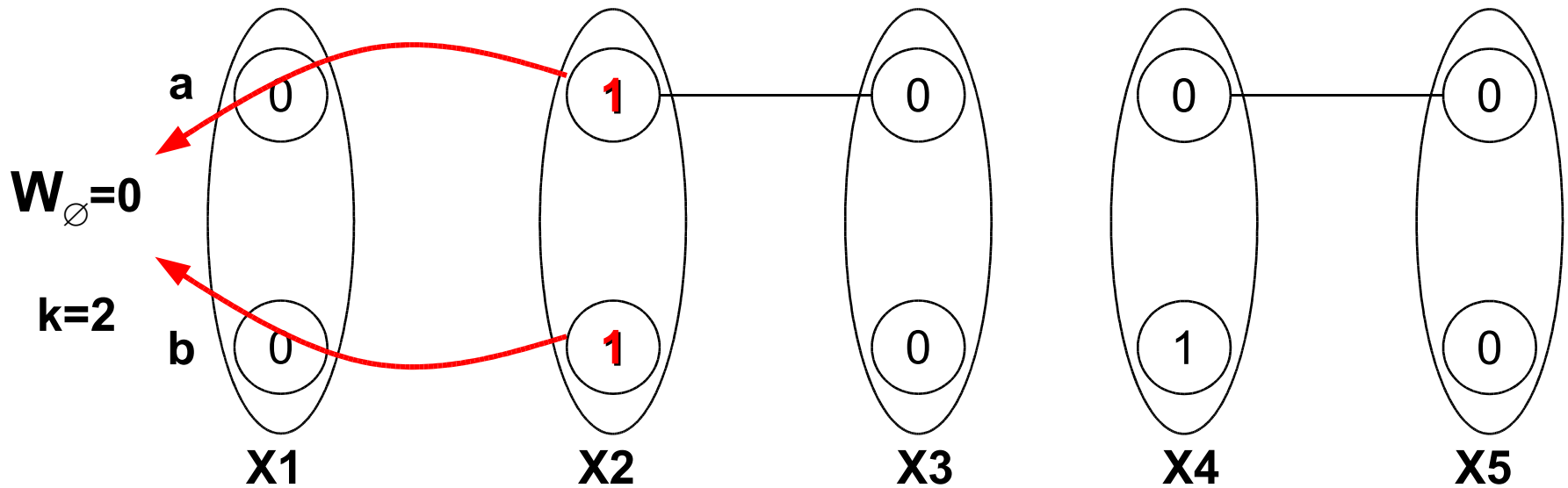
**Two cost projections  
from two binary constraints  
to a unary constraint**

# Soft local consistency



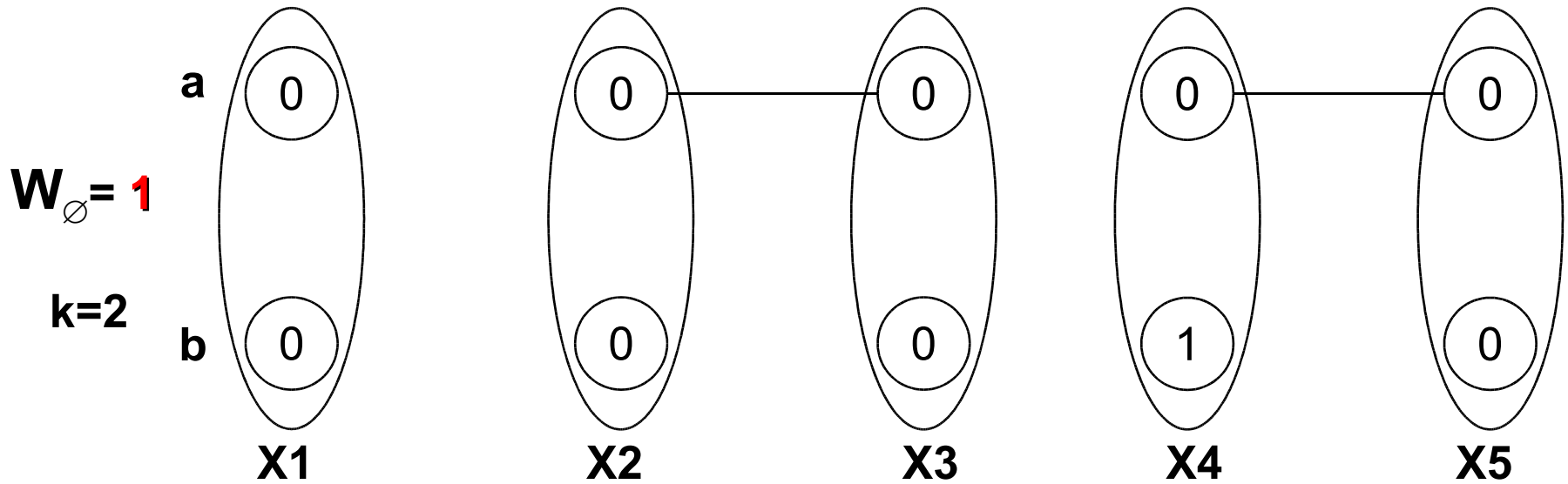
**Two cost projections  
from two binary constraints  
to a unary constraint**

# Soft local consistency



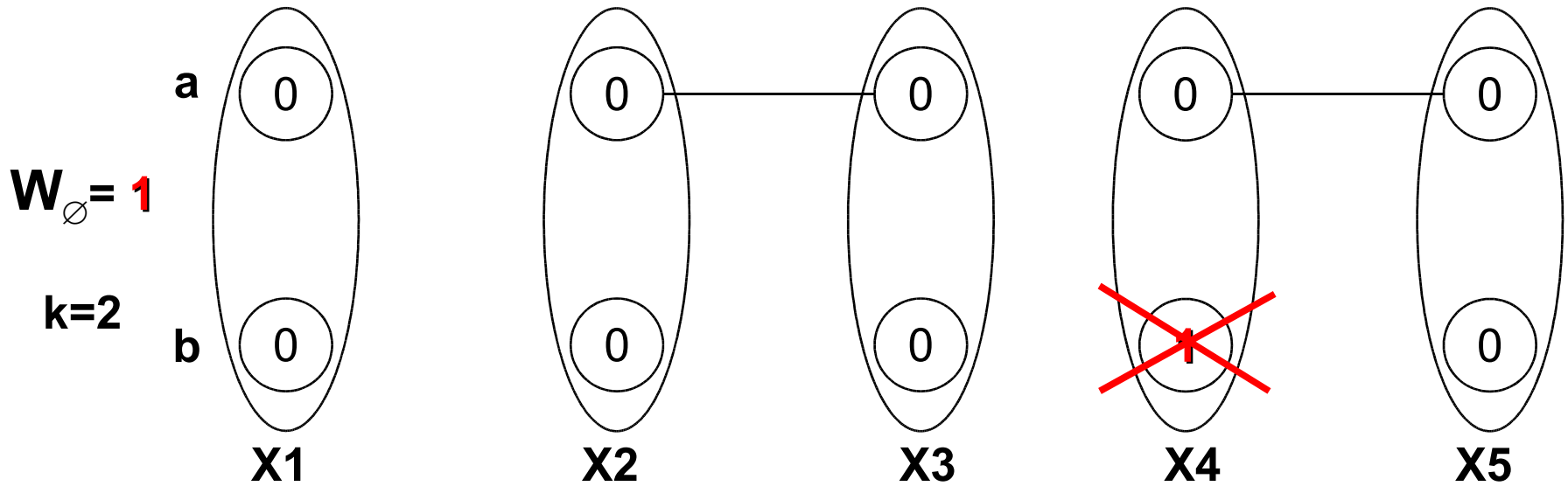
**Cost projection  
from a unary to the  
zero-arity constraint**

# Soft local consistency



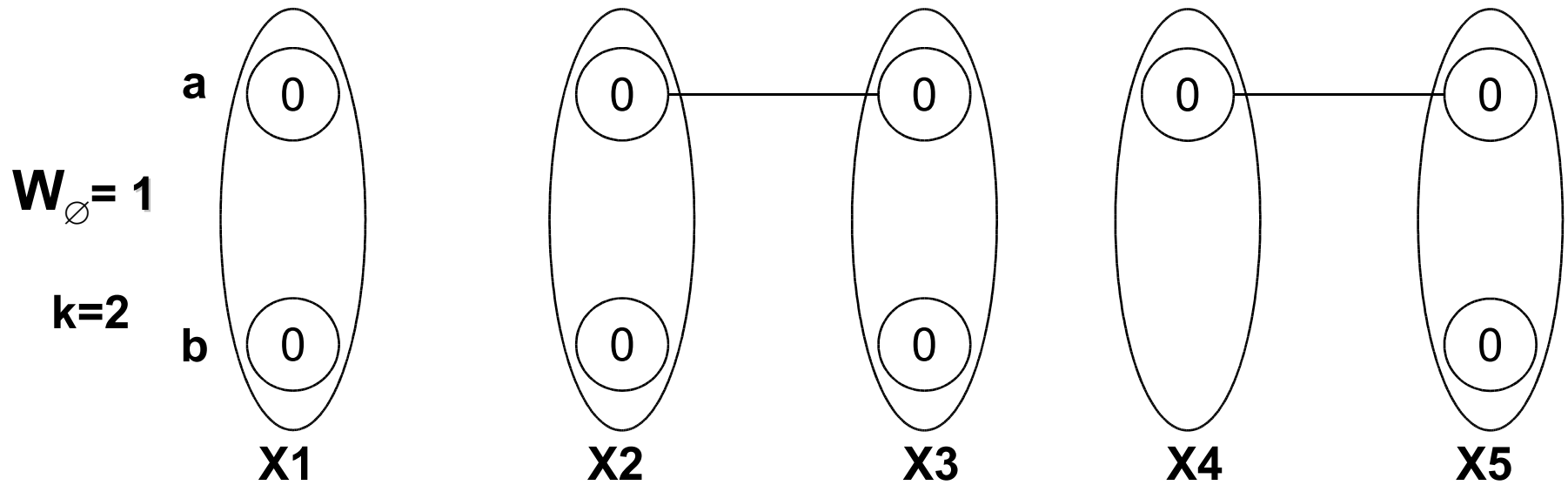
**Cost projection  
from a unary to the  
zero-arity constraint**

# Soft local consistency



**Value removal**

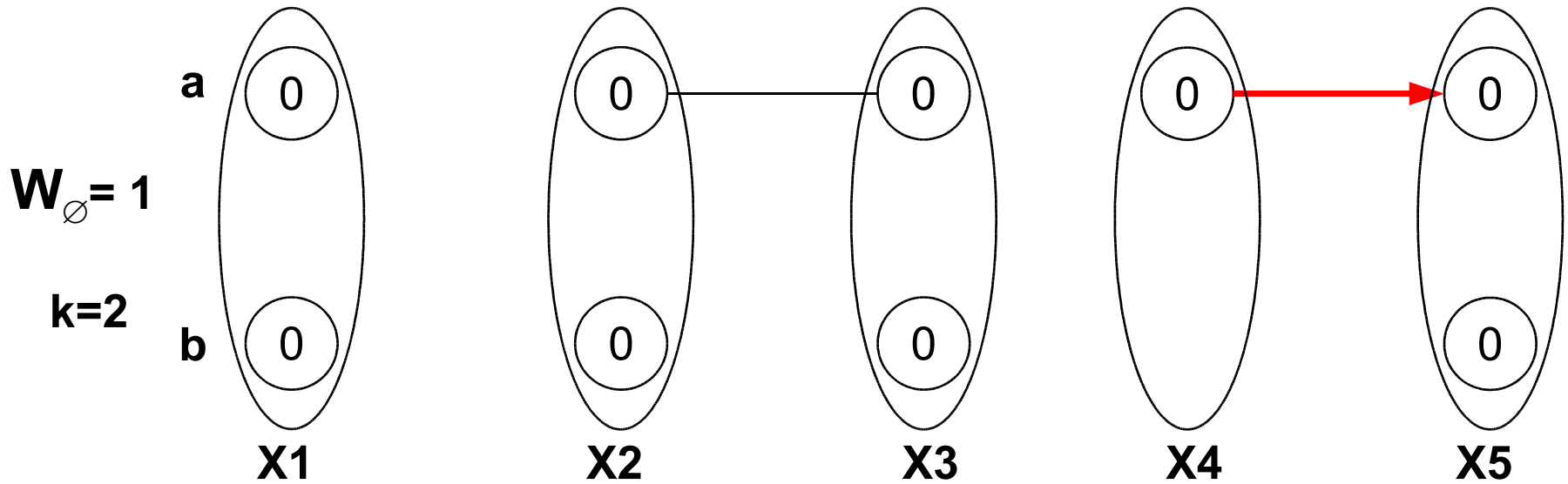
# Soft local consistency



**Value removal**

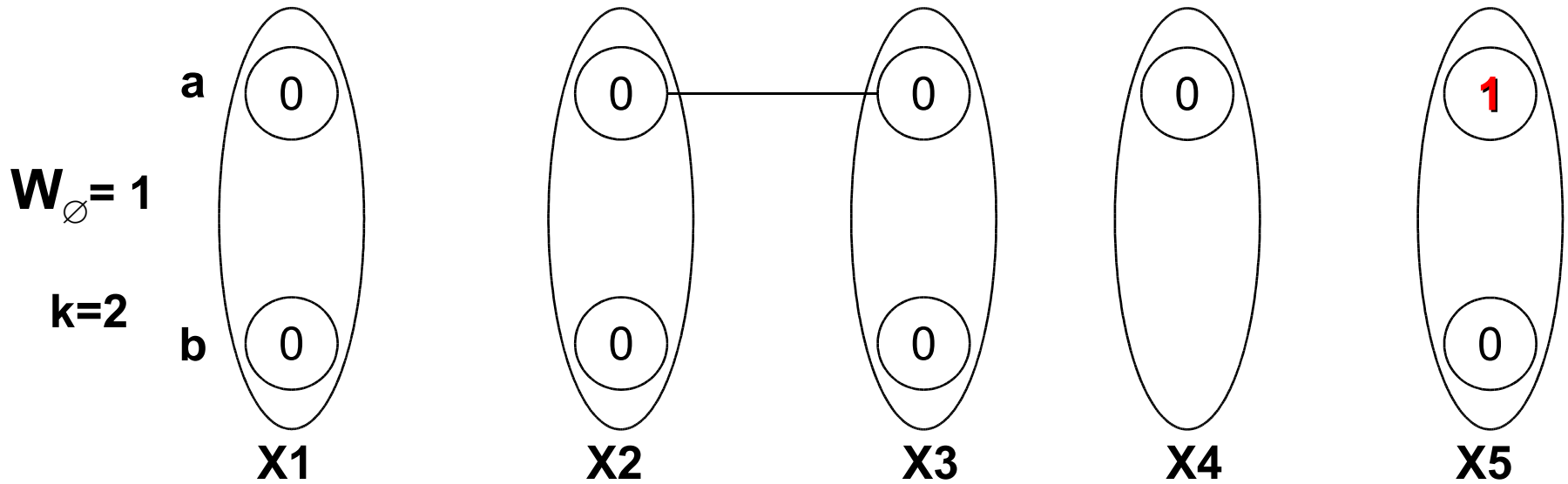


# Soft local consistency



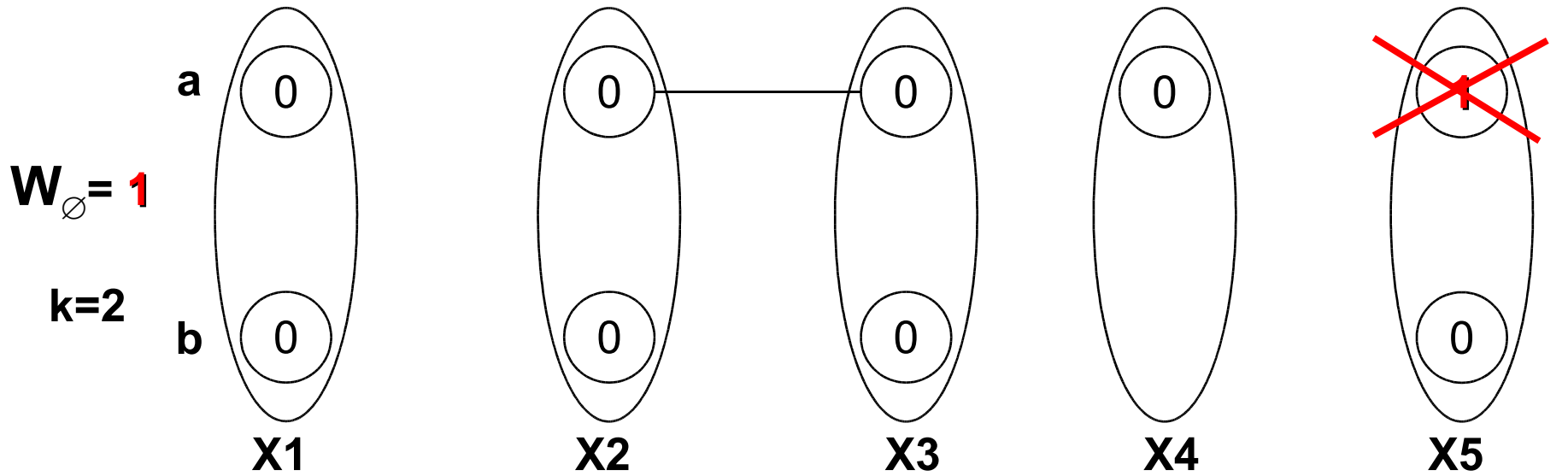
**Cost projection  
from a binary to  
a unary constraint**

# Soft local consistency



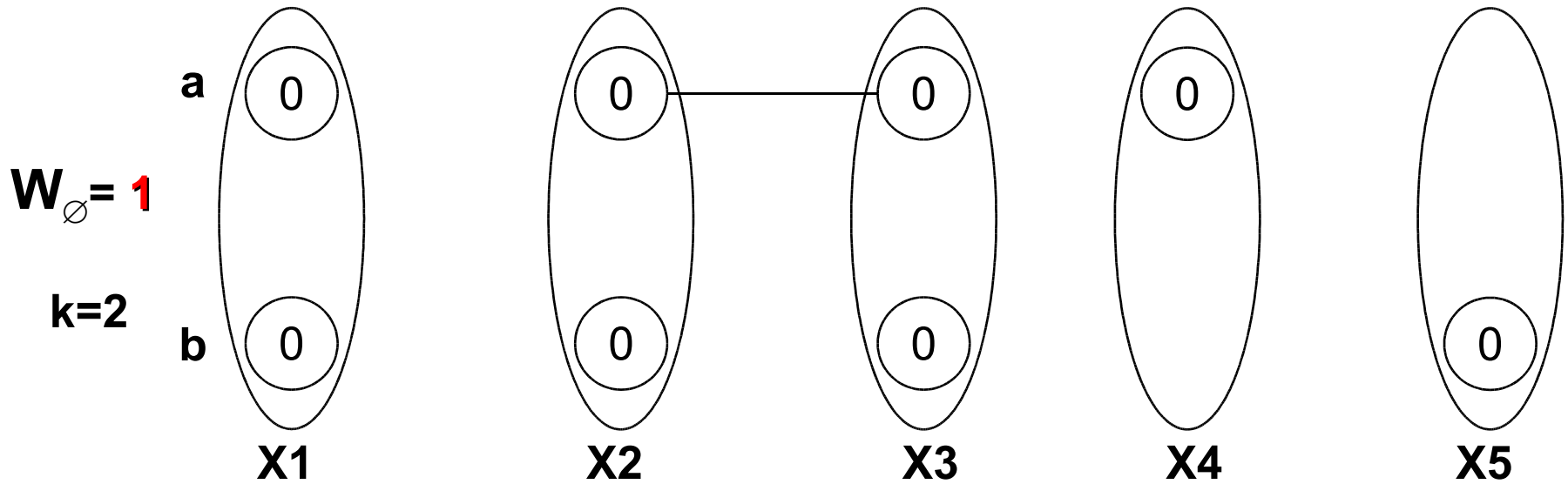
**Cost projection  
from a binary to  
a unary constraint**

# Soft local consistency



**Value removal**

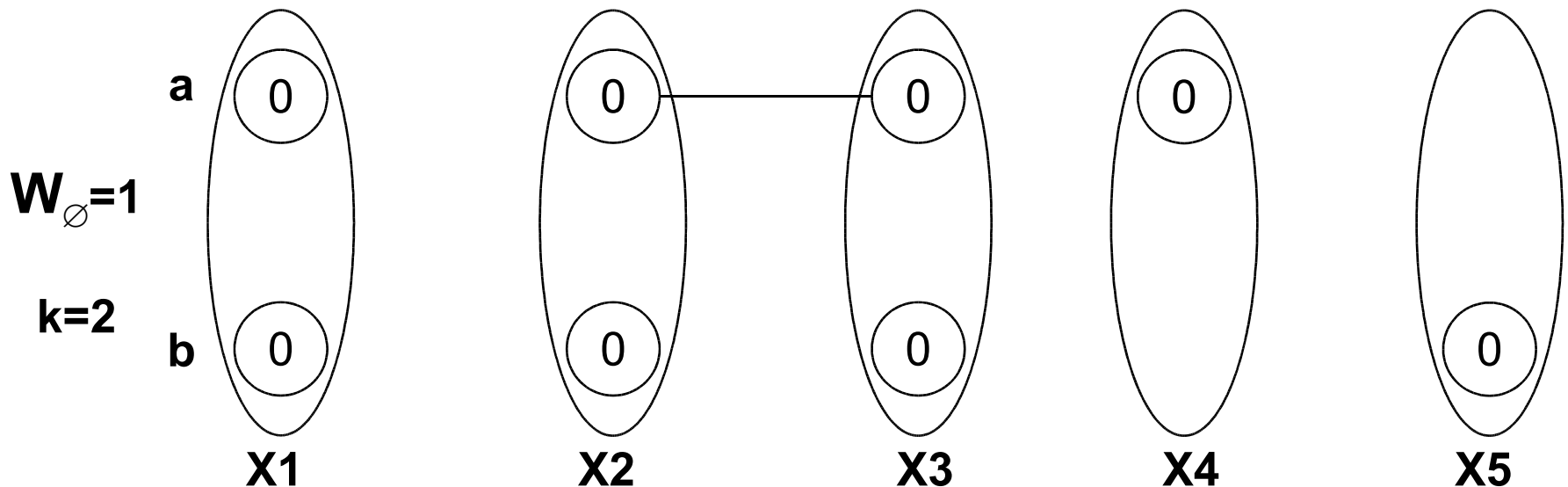
# Soft local consistency



**Value removal**

# Soft local consistency: **various levels**

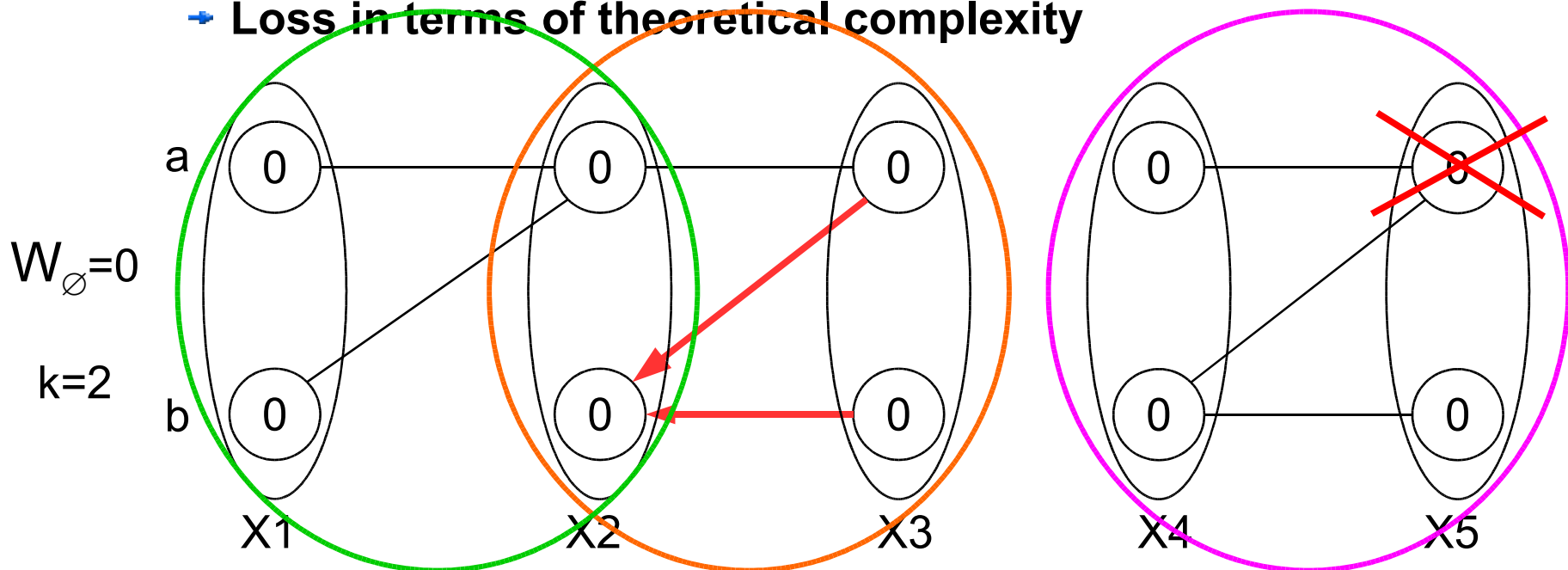
- **NC\***: Node Consistency
- **AC\***: Arc Consistency
- **DAC\***: Directed Arc Consistency
- **FDAC\***: Full Directed Arc Consistency



# Tree decomposition and soft local consistency

Two main difficulties:

- **Costs are moving between clusters and towards  $W_\emptyset$** 
  - Recorded subproblem lower bounds may be no longer valid
- **Value removals may affect any cluster**
  - No guarantee to improve the lower bound when revisiting the same subproblem
  - **Loss in terms of theoretical complexity**



# Three approaches considered

- **Limited form of soft local consistency** (forward-checking)
  - **FC-BTD** (Time:  $O(d^w)$ , Space:  $O(d^s)$ )
- **Limited soft arc consistency**, with corrected recorded lower bounds and value removals limited to the current subproblem
  - **FDAC-BTD+** (Time:  $O(kd^w)$ , Space:  $O(d^s)$ )
- **Unlimited soft arc consistency**, without learning
  - **FDAC-PTS** (Time:  $O(d^h)$ , Space:  $O(nd)$ )

# Experimental results

## Radio Link Frequency Allocation Problem

RLFAP optimum $n, d, w, h$	SUB <sub>1</sub>		SUB <sub>4</sub>		SCEN-06	
	2669		3230		3389	
	14, 44, 13, 14		22, 44, 19, 21		100, 44, 19, 67	
Method	time	#LB	time	#LB	time	#LB
FC-BTD	1197	0	-	0	-	-
NC-BTD+	490	0	-	0	-	-
FDAC-BTD+	14	0	929	0	10,309	326
FDAC-PTS	14	<i>n/a</i>	851	<i>n/a</i>	-	<i>n/a</i>
MFDAC	14	<i>n/a</i>	984	<i>n/a</i>	-	<i>n/a</i>

**The first time the whole SCEN-06 instance  
is solved by a search algorithm**

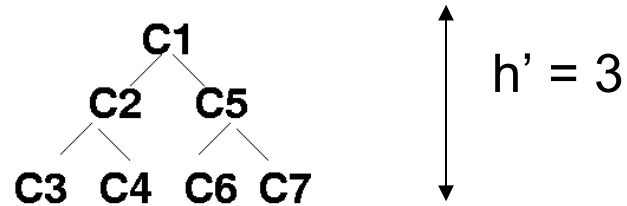
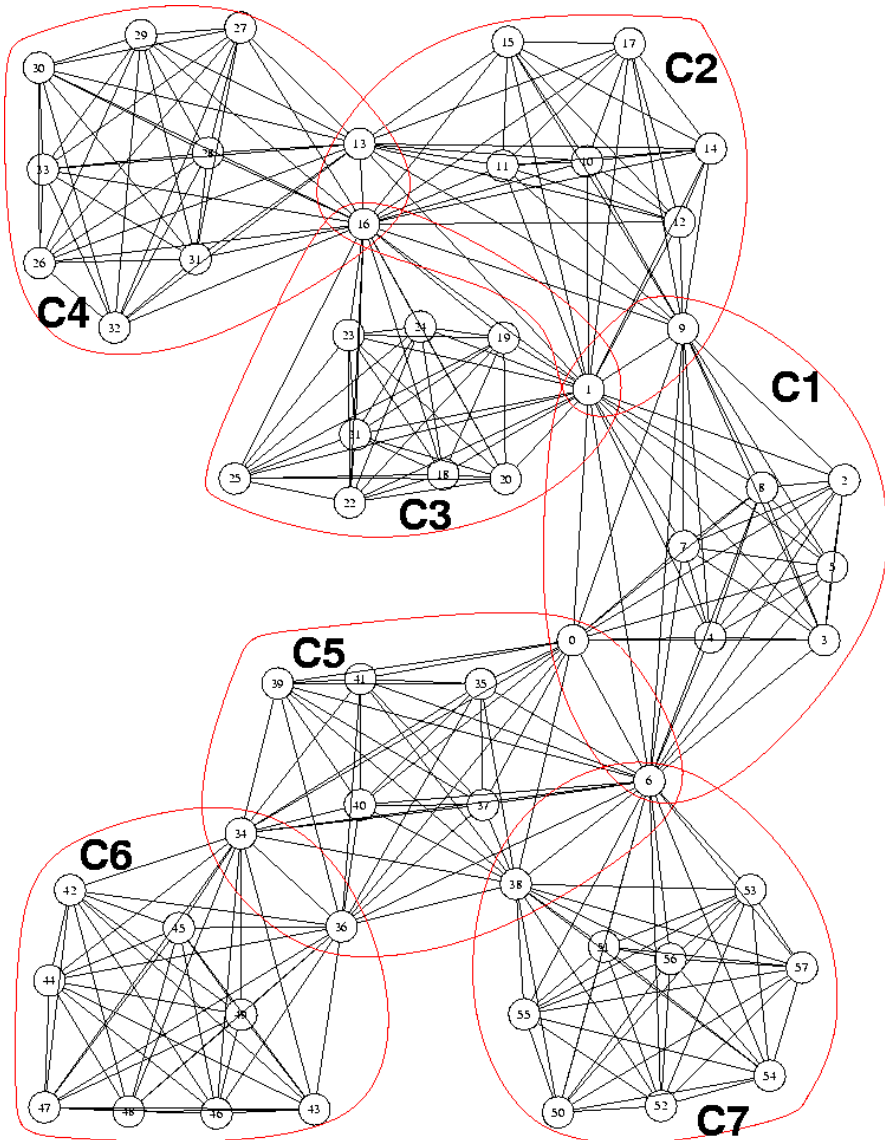
**To be observed: small amount of memory required**



# Conclusion

- **FDAC-BTD+**:
  - Cluster tree problem decomposition
  - Tree search
  - Graph-based backjumping and learning
  - (Limited) soft arc consistency enforcing
  - Initial upper-bounds
- **Cluster tree decomposition** and **soft local consistency** can be combined, but various technical options can be considered, and must be more widely experimented.

# Random Binary Trees of Cliques



## Parameters

Clique size  $(w+1) = 10$

**Separator size (s)**

Clique tree height  $(h') = 3$

Domain size  $(d) = 5$

**Constraint tightness**

Tree decomposition based on  
Maximum Cardinality Search

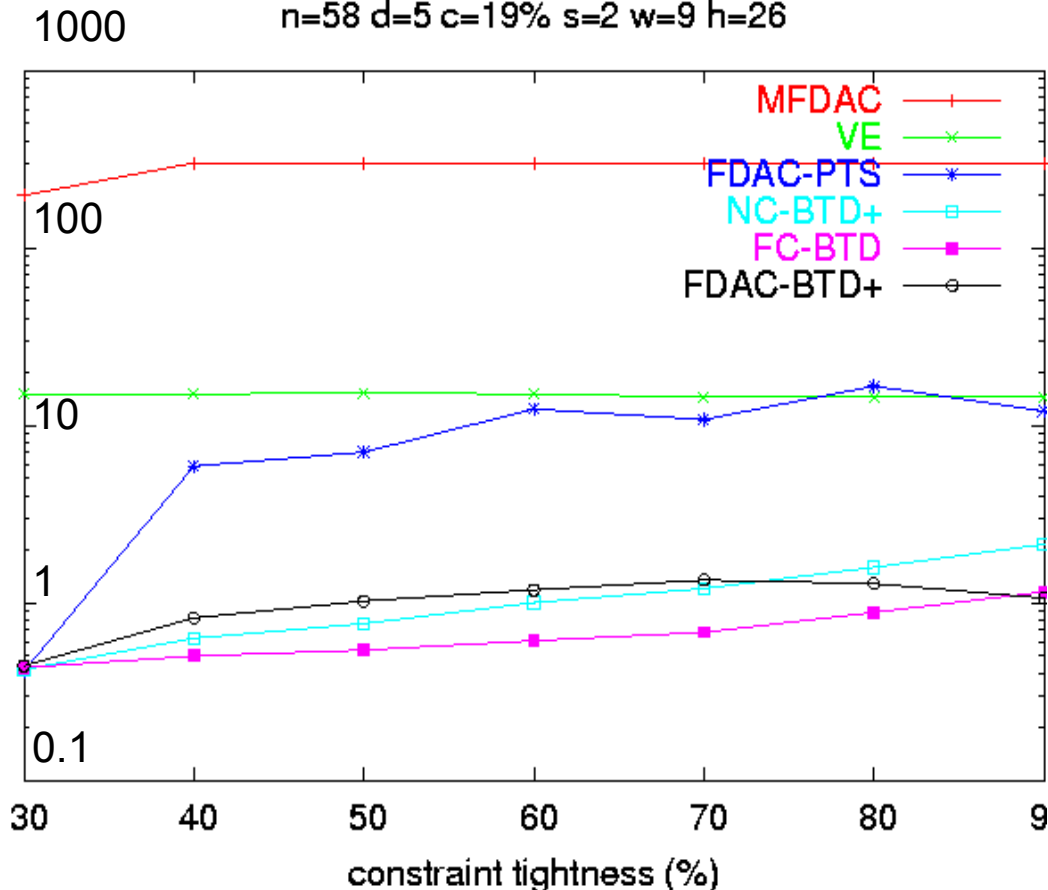
Root selection minimizing  
tree-height (h)

# Random Binary Trees of Cliques

**Small-size separator  $s = 2$**

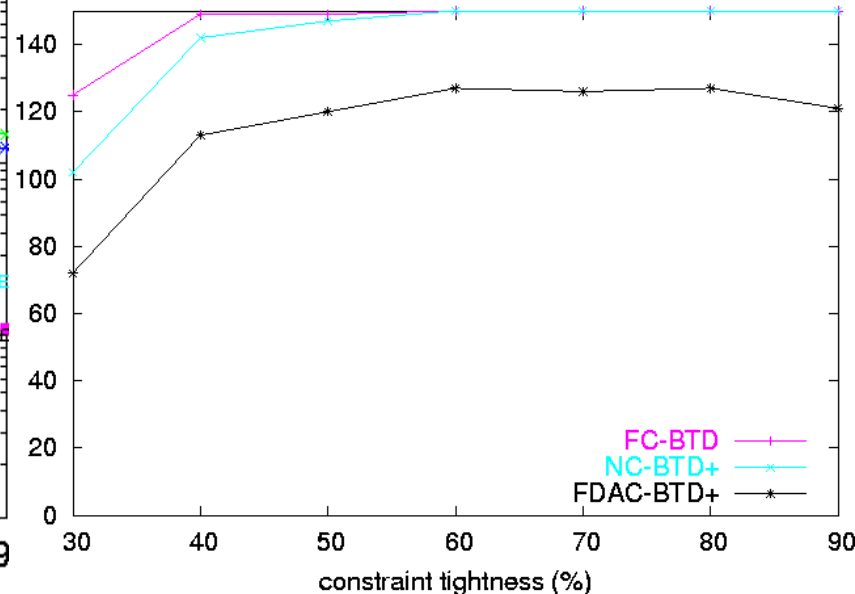
CPU-time (sec.)

$n=58$   $d=5$   $c=19\%$   $s=2$   $w=9$   $h=26$



Memory space (#lb)

$n=58$   $d=5$   $c=19\%$   $s=2$   $w=9$   $h=26$



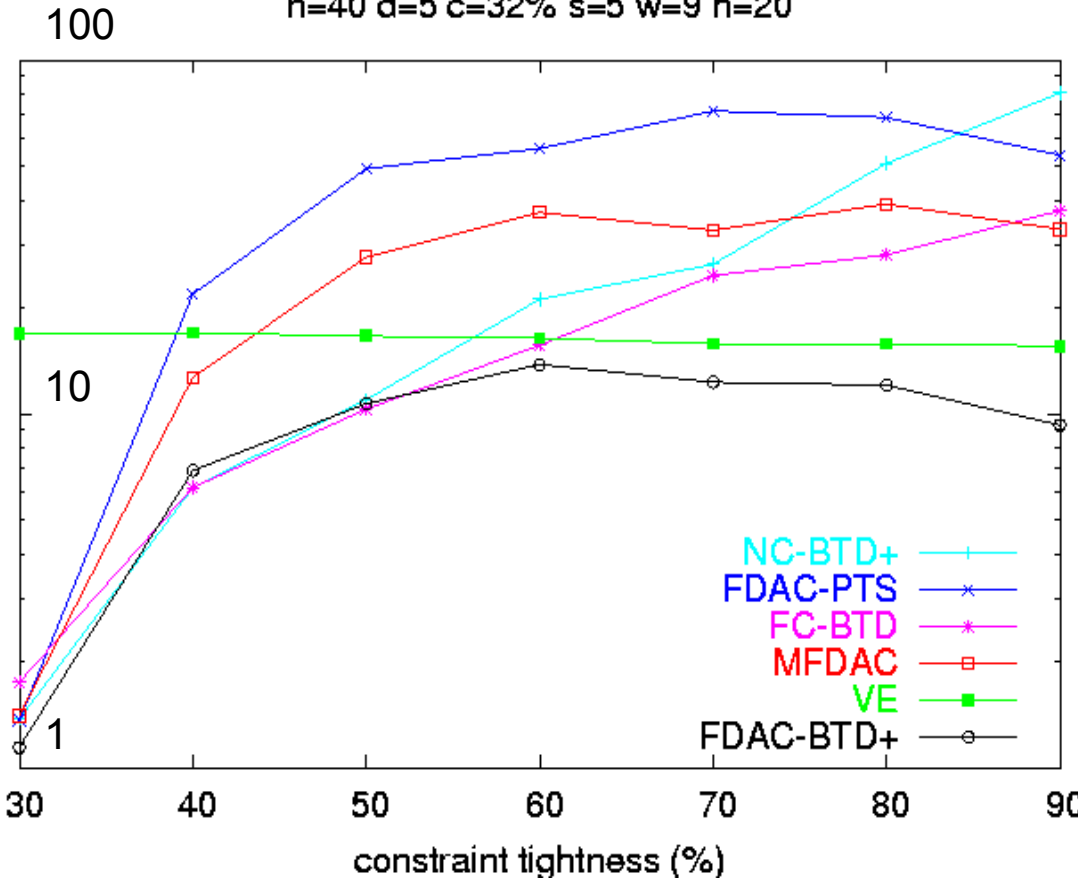
dom/deg dynamic variable ordering. CPU-time to find the optimum and prove its optimality

# Random Binary Trees of Cliques

**Medium-size separator  $s = 5$**

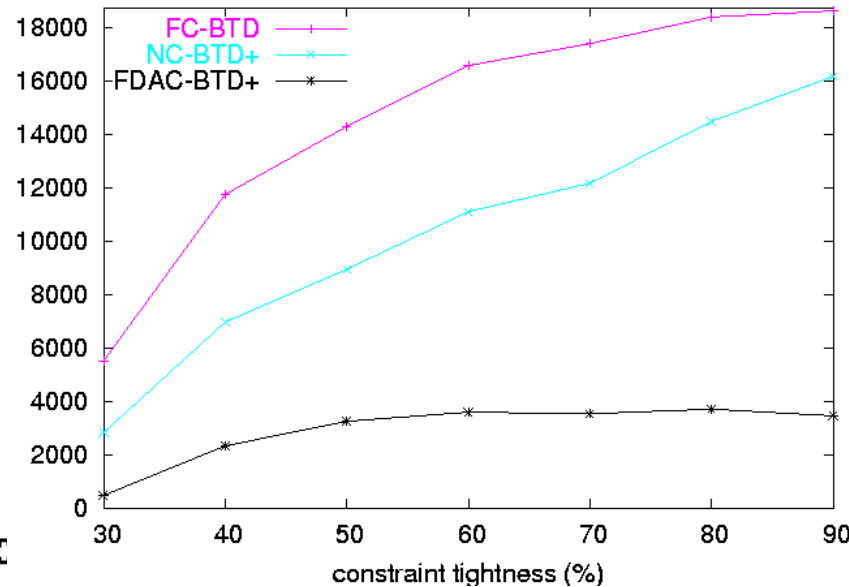
CPU-time (sec.)

$n=40$   $d=5$   $c=32\%$   $s=5$   $w=9$   $h=20$



Memory space (#lb)

$n=40$   $d=5$   $c=32\%$   $s=5$   $w=9$   $h=20$



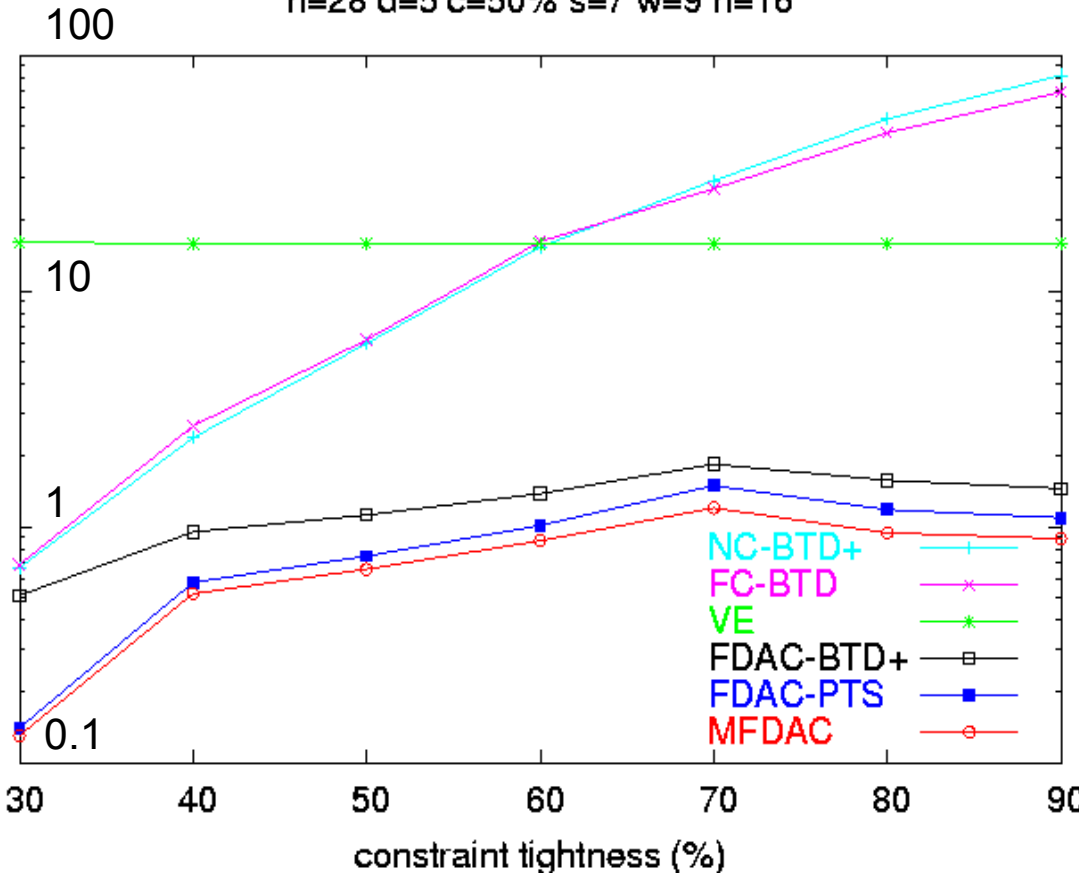
dom/deg dynamic variable ordering. CPU-time to find the optimum and prove its optimality

# Random Binary Trees of Cliques

**Large-size separator  $s = 7$**

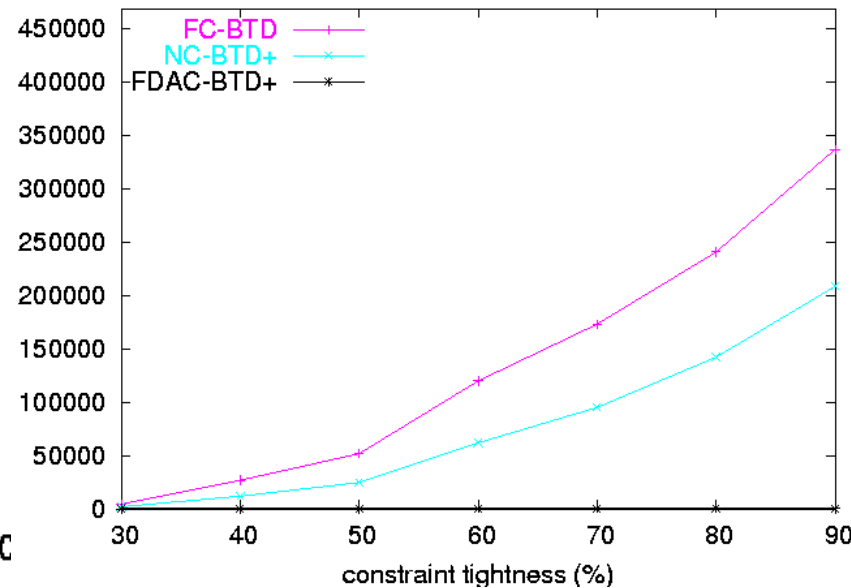
CPU-time (sec.)

$n=28$   $d=5$   $c=50\%$   $s=7$   $w=9$   $h=16$



Memory space (#lb)

$n=28$   $d=5$   $c=50\%$   $s=7$   $w=9$   $h=16$



dom/deg dynamic variable ordering. CPU-time to find the optimum and prove its optimality

