



# Effective Bounding Techniques For Solving Unate and Binate Covering Problems

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## Related Work

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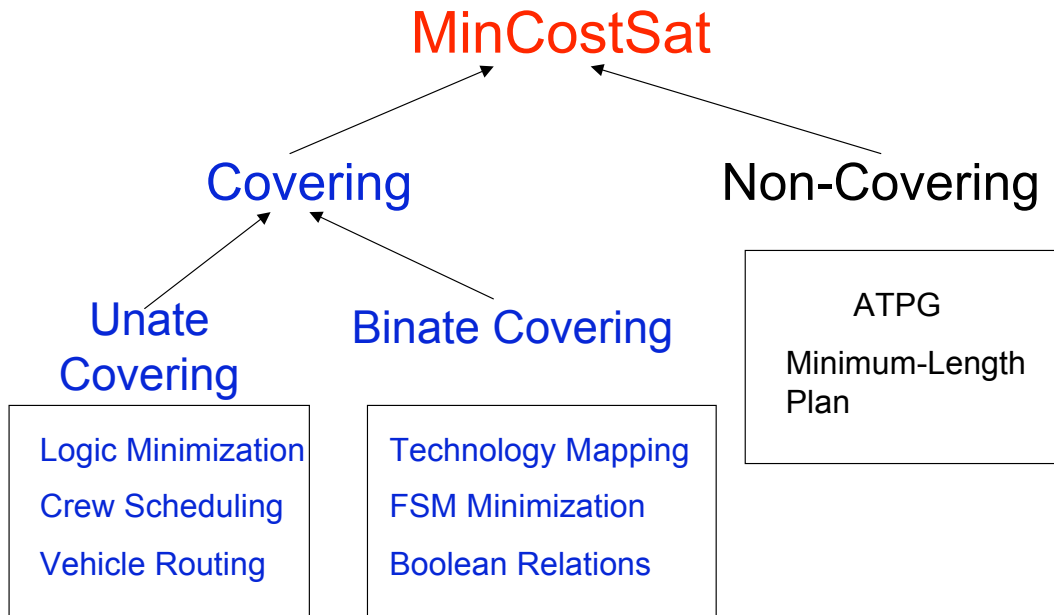
- Unate (Set) Cover Problem (UCP)  
1987: *espresso*, Rudell et al  
1993: *espresso-signature*, McGeer et al  
....
- Binate Cover Problem (BCP)  
1996: *scherzo*, Coudert  
2002: *bsolo*, Manquinho et al  
...

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# Native MinCostSat Problems

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## Outline

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1. Background
  2. Our UCP/BCP Solver
  3. Experimental Results
  4. Conclusions
- 

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## Unate (Set) Covering

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Three backup positions to fill.  
 Five players to choose from.  
 Want to minimize the number of players.



Duncan



Francis



Malone



Stockton



Yao

Guard		1		1	
Forward	1		1		
Center	1		1		1

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## Unate Covering as CNF Sat

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- Conjunctive Normal Form

**Guard**



$F \vee S$

**Forward**



$D \vee M$

**Center**



$D \vee M \vee Y$

- Goal: minimize  $F + S + D + M + Y$

Min-cost solutions:  $F = 1, D = 1, S = M = Y = 0$

$M = 1, S = 1, D = F = Y = 0$

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## Binate Covering Problem

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Additional constraint 1:

Malone and Stockton

are together ( $M \Leftrightarrow S$ )

$$\overline{M} \vee S$$

$$M \vee \overline{S}$$

Additional constraint 2:

Duncan and Stockton

can't be signed together ( $\overline{D} \wedge \overline{S}$ )

$$\overline{D} \vee \overline{S}$$

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## Binate Covering Problem (cont.)

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**constraints**

$$F \vee S$$

$$D \vee M$$

$$D \vee M \vee Y$$

$$\overline{M} \vee S$$

$$M \vee \overline{S}$$

$$\overline{D} \vee \overline{S}$$



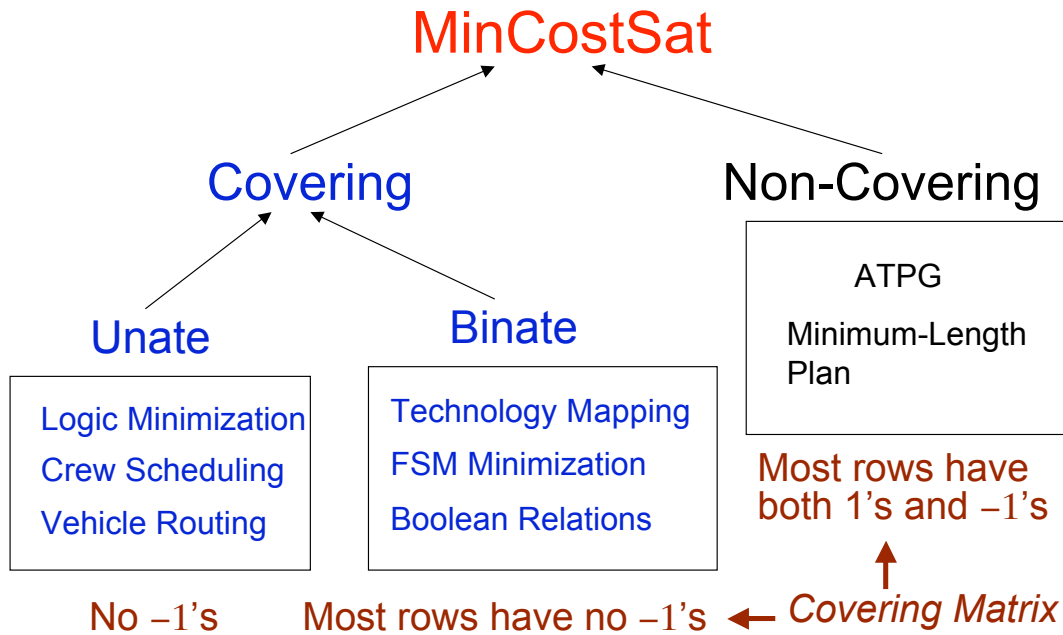
**covering matrix**

	D	F	M	S	Y
		1		1	
	1		1		
	1		1		1
			-1	1	
			1	-1	
	-1			-1	

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# MinCostSat Problems Revisited



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## Solve MinCostSat as ILP

MinCostSat is a special case of 0-1 integer linear programming

For each constraint in the MinCostSat instance...

$$\bar{x}_1 \vee x_2 \vee \bar{x}_3 \vee x_4$$

...replace  $\bar{x}_i$  with  $(1 - x_i)$ :

$$(1 - x_1) + x_2 + (1 - x_3) + x_4 \geq 1$$

$$-x_1 + x_2 - x_3 + x_4 \geq -1$$

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# MinCostSat as ILP: an example

MinCostSat

ILP

$$\bar{x}_1 \vee x_2 \vee \bar{x}_3 \vee x_4$$

$$-x_1 + x_2 - x_3 + x_4 \geq -1$$

$$\bar{x}_1 \vee \bar{x}_2 \vee \bar{x}_4$$



$$-x_1 - x_2 + x_4 \geq -2$$




$$x_1 \vee x_2 \vee x_3$$

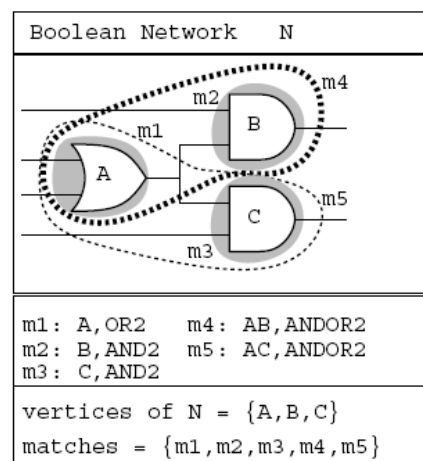
$$x_1 + x_2 + x_3 \geq 1$$

Assume unit cost, minimize:  $x_1 + x_2 + x_3 + x_4$

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# Technology Mapping

Library L	Cost
 AND2	4
 OR2	4
 ANDOR2	5



Cover A  $m_1 \vee m_4 \vee m_5$

Cover B  $m_2 \vee m_4$

Cover C  $m_3 \vee m_5$

$m_2 \rightarrow m_1$   $\bar{m}_2 \vee m_1$

$m_3 \rightarrow m_1$   $\bar{m}_3 \vee m_1$

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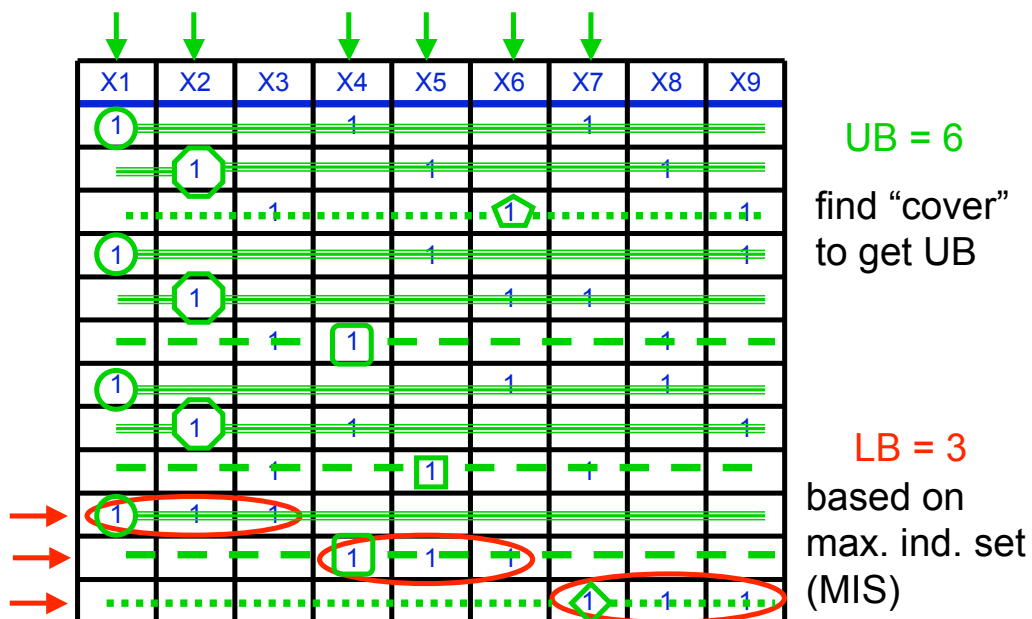
# Our UCP/BCP Solver

## Classical branch and bound

1. **Upper/lower bound** calculation  
(for a unate example)
2. **Branching** (variable assignment)
3. One iteration of the main algorithm
4. Effect of better (global) upper bounds  
and (local) lower bounds.

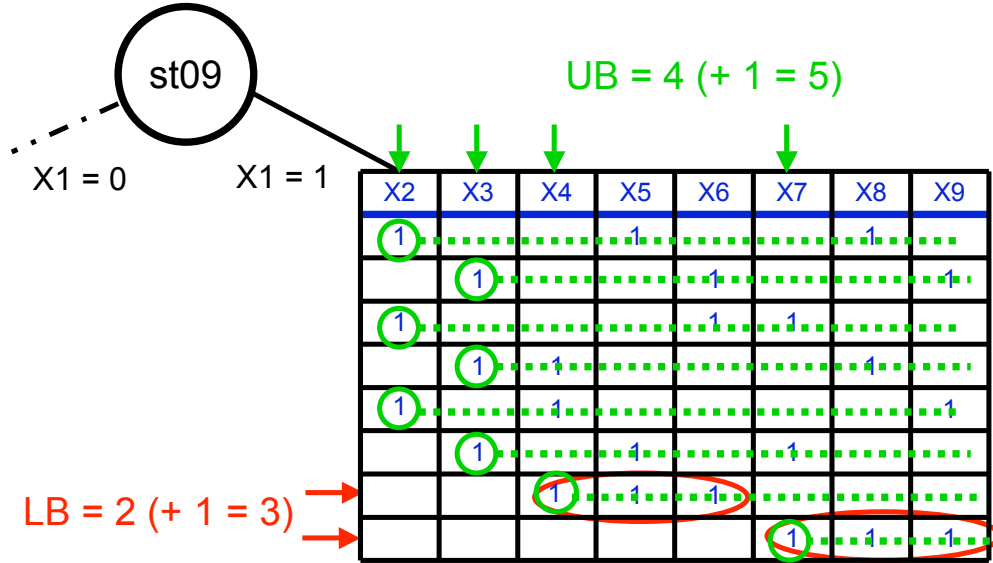
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## An Example (Steiner 9)



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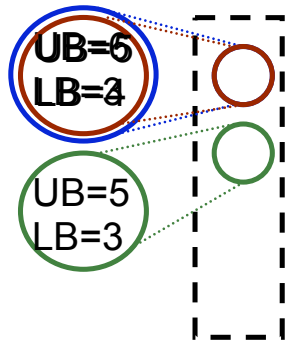
## After assignment of a variable



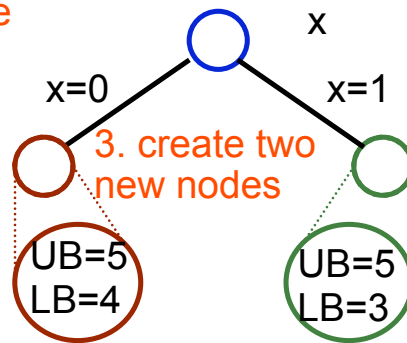
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## The algorithm: one iteration

1. dequeue "best" node in priority queue



2. select a variable

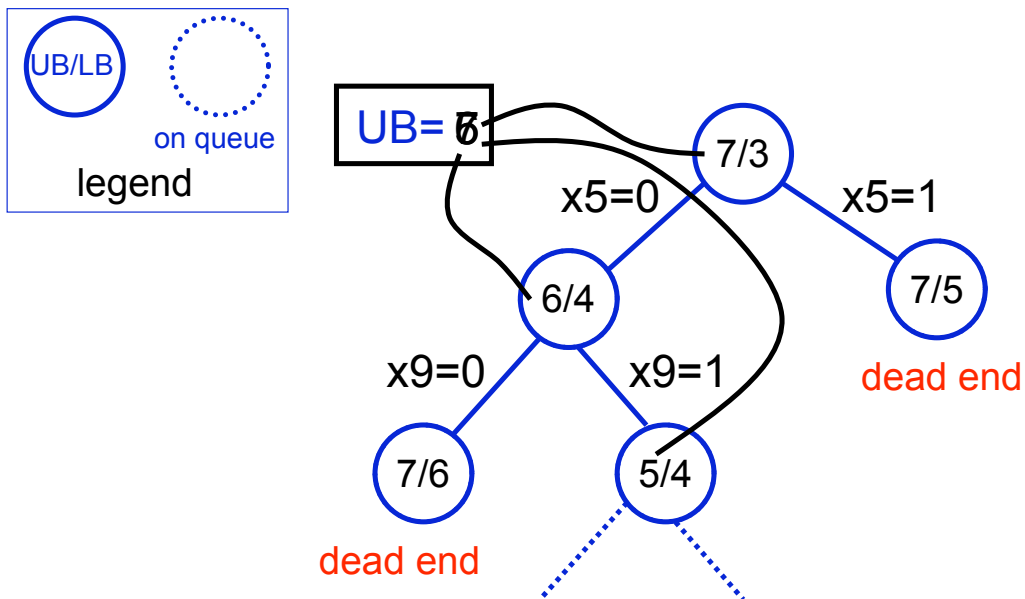


4. do reduction and compute bounds

5. enqueue both new nodes

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## Lower and Upper Bounds



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## Branch-and-bound solver: eclipse

### The algorithm

Initialize root; Q.enqueue(root);  $globalUB = UB(\text{root})$

```
while ( Q is not empty ) {  
  node = Q.dequeue();  
  if ( LB(node) < globalUB ) {  
    select branching variable  $x_i$   
    for ( b = 0 to 1 ) {  
       $n_b = (\text{node for } x_i = b)$   
      do reduction and bounds for  $n_b$   
      if ( UB( $n_b$ ) < globalUB ) { globalUB = UB( $n_b$ ) }  
      if ( LB( $n_b$ ) < globalUB ) { Q.enqueue(  $n_b$  ) }  
    }  
  }  
}
```

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# Eclipse Performance Factors

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- Seven Performance Factors
  1. Lower Bounding (ILP techniques)
  2. Upper Bounding (stochastic search)
  3. Search-Tree Exploration Strategies
  4. Branching Variable Selection
  5. Search Pruning
  6. Reductions
  7. Data Structures

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## Factor 1: Lower Bounding

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Three lower-bounding strategies:

- Maximum Independent Set (MIS)
- Linear Programming Relaxation (LPR)
- Cutting Planes (CP)

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## Lower Bounding: MIS

- MIS - maximum independent set of rows

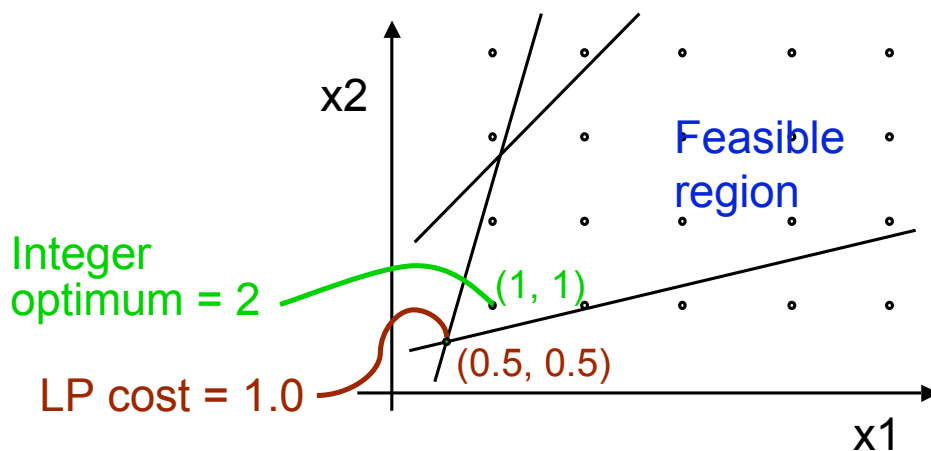
	x1	x2	x3	x4	x5	x6
row1	1	1				
row2			1	1	1	
row3			1	1		
row4					1	1
row5			-1		-1	1

- Row 1, 2 form an independent set  $\Rightarrow$  LB = 2
- Row 1, 3, 4 form an independent set  $\Rightarrow$  LB = 3

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## Lower Bounding: LP Relaxation

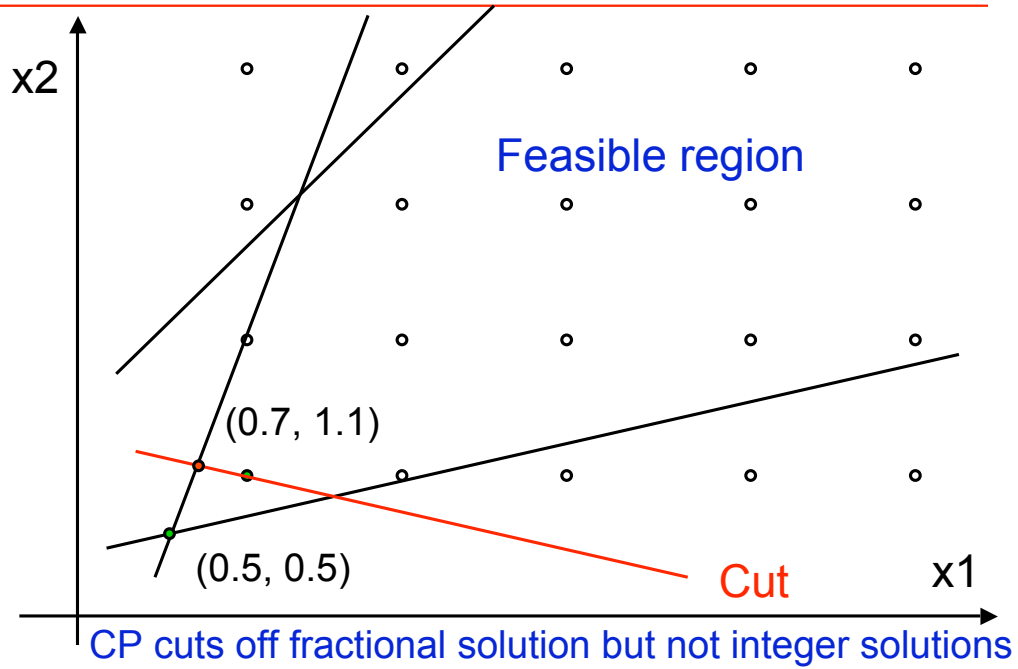
Relax integer constraints (LPR):



LP optimum is a lower bound on integer optimum

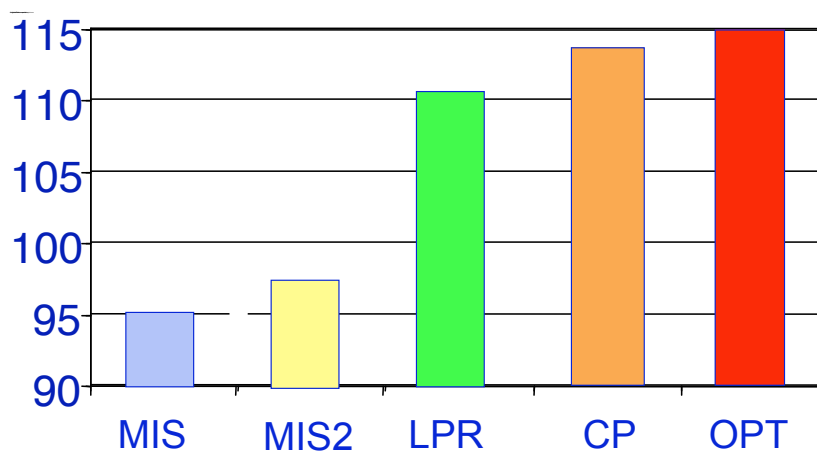
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## Lower Bounding: Cutting Planes (CP)



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## Lower Bound Methods (rot.b)



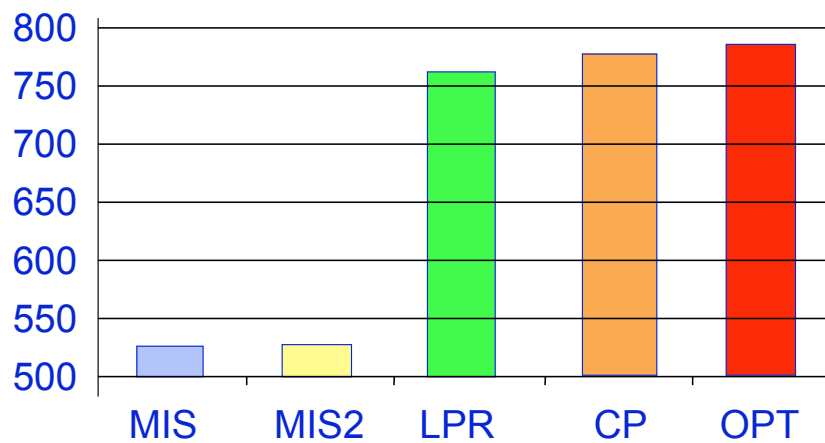
(MIS2 is MIS with extra tricks)

Most effective by far: CP

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## Lower Bound Methods (apex4.a)

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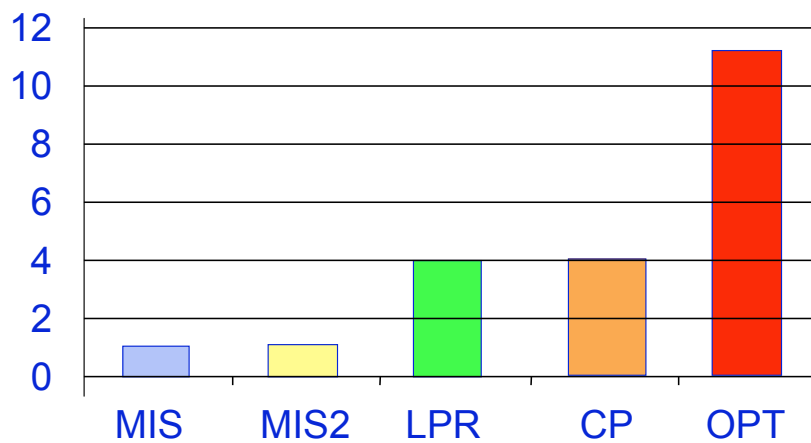
Most effective: CP

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## Lower Bound Methods (c1908\_F)

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no bounding method appears effective here ...  
(not a case of covering problem instance)

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## Next Performance Factor:

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1. Lower Bounding (ILP techniques)
- 2. Upper Bounding (stochastic search)**
3. Search-Tree Exploration Strategies
4. Branching Variable Selection
5. Search Pruning
6. Reductions
7. Data Structures

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## Factor 2: Upper Bounding

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- Stochastic local search for optimal solution:
  1. Apply at the **root** only or **at each node**
  2. Initialization – solution based on linear programming for lower bound
- For comparison purposes we initialize the global UB to the cost of the optimal solution (**oracle**)

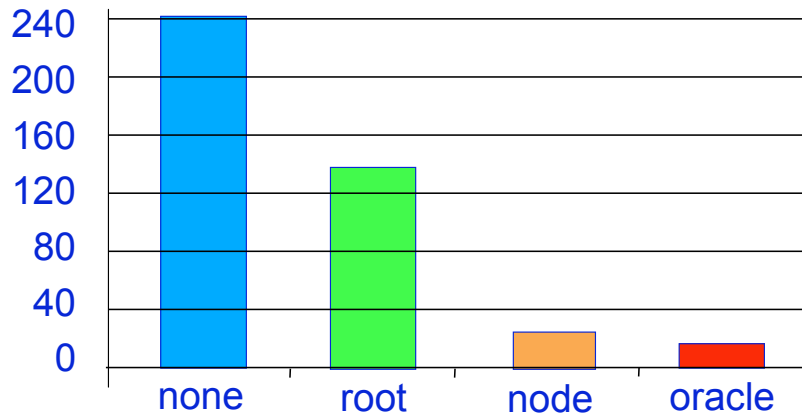
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## Upper Bound Methods (max1024)

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we report average cpu-seconds to solve the instance with different UB strategies using the same LB (the best possible, i.e. CP)



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## Local Search Initialization

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... random initialization of local search is not effective, see the table of runtimes (in cpu-seconds):

benchmark	random	rounded LP solution
ex5	16.3	15.1
exam.pi	20.8	5.4
max1024	143.9	27.5
bench1.pi	300**	4.7

\*\*timeout value

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## Factor 3: Tree-Exploration Strategy

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... search using priority queue performs better than depth-first search (recursion - the usual method):

(runtimes in cpu-seconds)

benchmark	depth-first	priority*
exam.pi	7.0	5.4
bench1.pi	7.9	4.7
ex5	16.2	15.1
max1024	74.1	27.5

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\* smallest lower bound first (to try to reduce UB quickly)

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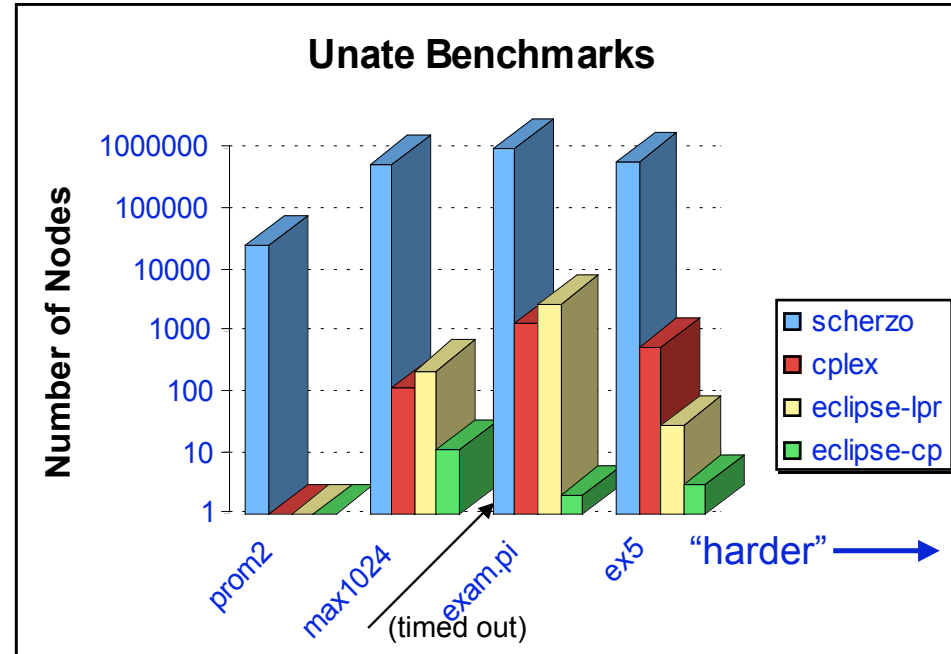
## Experimental Results

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- The solvers
    1. Scherzo [Coudert, DAC1996]
    2. Cplex (State of the art IP/ILP Solver)
    3. Eclipse-lpr (LP Relaxation)
    4. Eclipse-cp (Cutting Planes)
  - The benchmarks
    - Logic minimization (unate covering)
    - FSM minimization (binate covering)
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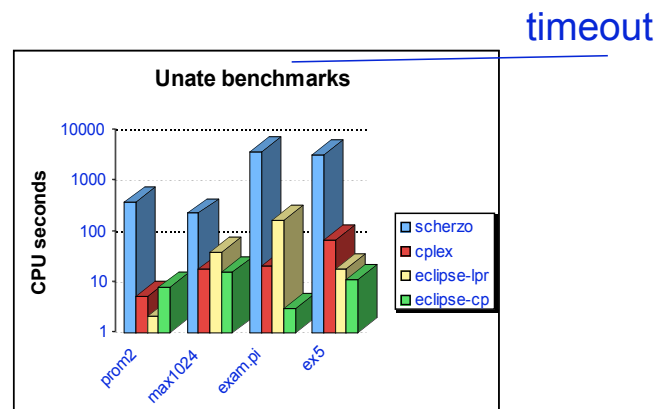
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# Unate Results (nodes visited)



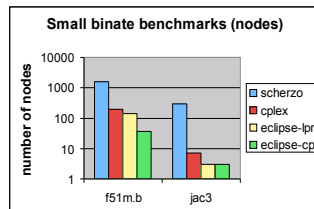
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# Typical Results: Unate (runtime)

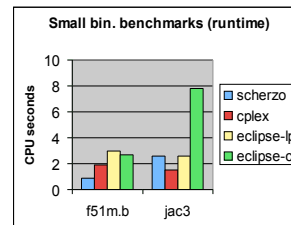


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# Typical Results: small binate

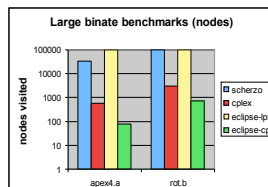


eclipse-cp spends a lot of time with cuts at each node (cplex visits more nodes and does cuts less frequently)



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# Large binate benchmarks

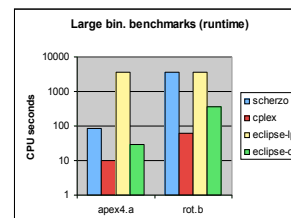


CPLEX only does cuts at some nodes.

We visit fewer nodes...

... but spend more time

We do cuts everywhere  
-- could be more judicious



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## latte = espresso – mincov + eclipse-cp

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benchmark	espresso (hours)		latte (secs)	
	cover	total	cover	total
prom2	---	12**	12.9	14.7
ex5	---	12**	129.5	139.2
max1024	---	12**	329.1	329.5

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\*\*timeout

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benchmark	prod-orig	prod-heur	prod-latte
prom2	287	287	287
ex5	256	74	65
max1024	1024	274	259

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## The Last Words ...

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### Acknowledgments:

- for the benchmarks
- for the solvers (*aura*, *bsolo*, *espresso*, *scherzo*)
- for constructive reviews

### Under construction:

- code tune-ups before the release of *eclipse*
  - additional experiments
    - for the journal submission
    - for comprehensive web-posting of results
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