

# Multistage Optimization with the help of Quantified Linear Programming



TECHNISCHE  
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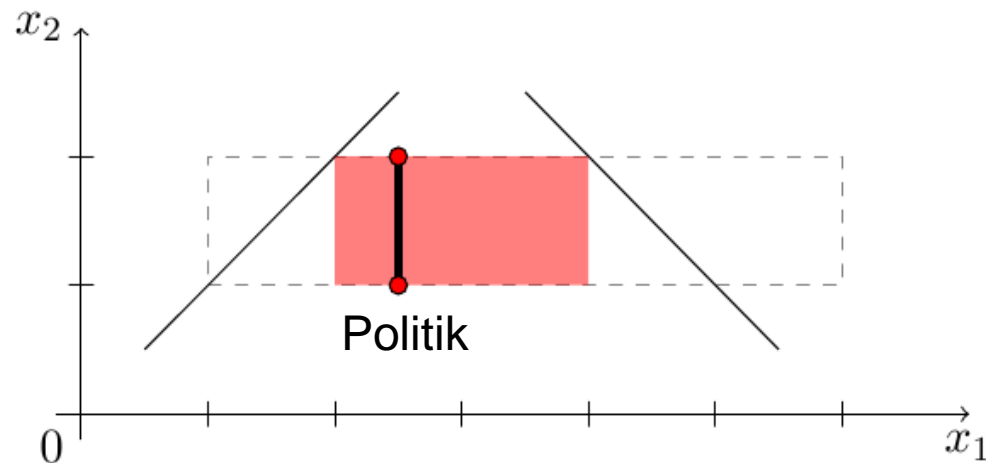
PD Dr. Ulf Lorenz

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# Quantified Linear Programs 1)

$$\exists x_1 \in [0, 1] \forall x_2 \in [0, 1] \exists x_3 \in [0, 1] : \\ \begin{pmatrix} 0 & -1 & -1 \\ -1 & 1 & 1 \\ 2 & 2 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \leq \begin{pmatrix} -1 \\ 1 \\ 3 \end{pmatrix}$$

- Union of all winning policies forms a polytope..



$$\exists x_1 \in [1, 6] \forall x_2 \in [1, 2] : x_1 + x_2 \leq 6 \wedge x_2 - x_1 \leq 0$$

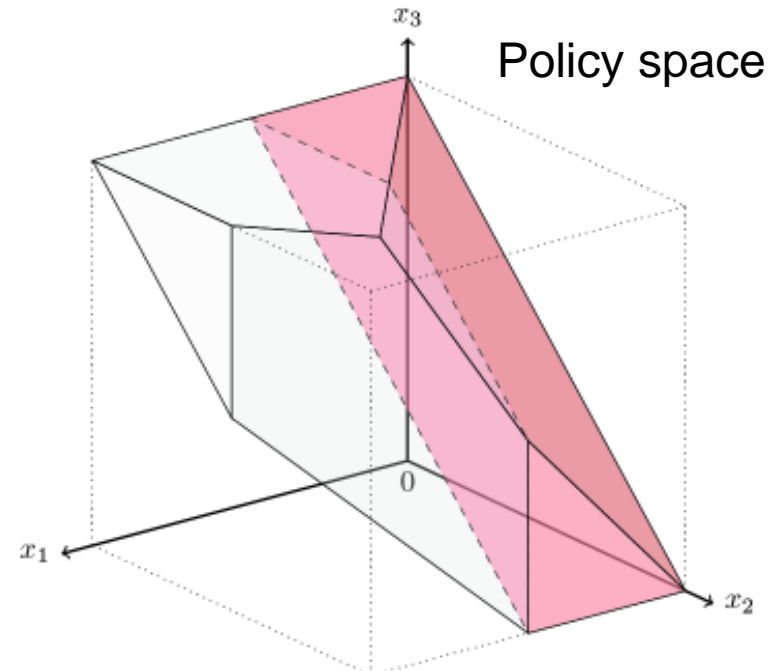
1) K. Subramani. Analyzing selected quantified integer programs. Springer, LNAI 3097, pages 342–356, 2004.

# Quantified Linear Programs

## Quantified Linear Program (QLP)

- ▶ continuous variables
- ▶ polyhedral solution space
- ▶ relaxation of QIP

Complexity is unknown in general.



- There is a winning policy against the universal player if and only if there is a winning policy against a universal player who is restricted to  $\{0,1\}$ .
- Vertex complexity stays good-natured as long as the number of quantifier changes is limited by a constant.

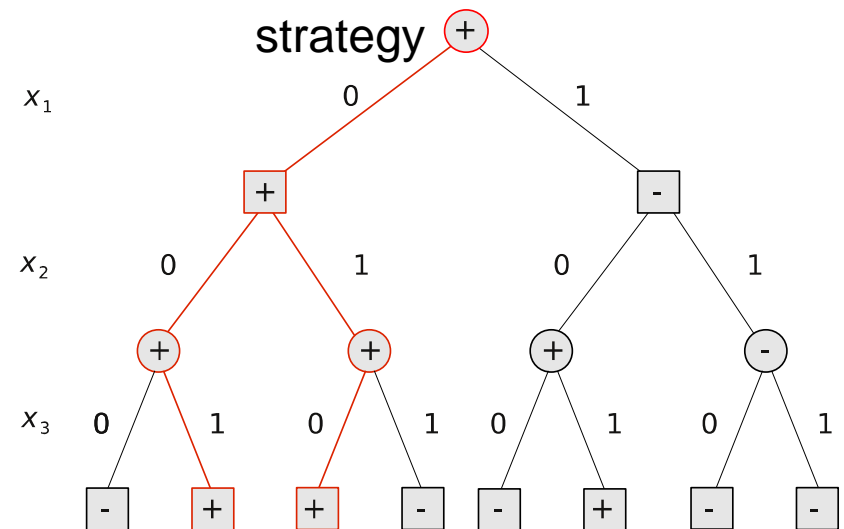
# Quantified Linear Integer Programs

$\exists x_1 \in [0, 1] \forall x_2 \in [0, 1] \exists x_3 \in [0, 1] , x \text{ integer:}$

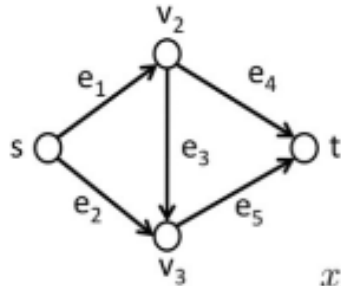
$$\begin{pmatrix} 0 & -1 & -1 \\ -1 & 1 & 1 \\ 2 & 2 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \leq \begin{pmatrix} -1 \\ 1 \\ 3 \end{pmatrix}$$

## Quantified Integer Program (QIP)

- ▶ integral variables
- ▶ can be visualized as a game-tree, strategy
- ▶ PSPACE-complete



# Example problem: Dynamic Graph Reliability



$\exists x_1, x_2 \forall y_{2,4}, y_{2,5} \exists x_3 \forall y_{3,4}, y_{3,5} \exists x_4, x_5, x_\Delta$  (all binary)

$$\left. \begin{array}{l} x_1 = x_4 + x_3 \\ x_2 + x_3 = x_5 \\ x_1 + x_2 = 1 \\ x_4 + x_5 = 1 \end{array} \right\} \text{Flow constraints}$$

$$\left. \begin{array}{l} x_4 \leq (1 - y_{2,4}) + (1 - x_1) + x_\Delta \\ x_4 \leq (1 - y_{3,4}) + (1 - x_2 - x_3) + x_\Delta \\ x_5 \leq (1 - y_{2,5}) + (1 - x_1) + x_\Delta \\ x_5 \leq (1 - y_{3,5}) + (1 - x_2 - x_3) + x_\Delta \end{array} \right\} \text{constraints for existential player}$$

$$2x_\Delta \leq y_{3,4} + y_{3,5} + y_{2,4} + y_{2,5} \left. \vphantom{2x_\Delta} \right\} \text{Critical constraint for the universal player}$$

## Further example problems

- Connect6
- system synthesis of a pump system
- two stage jobshop
- two stage car-sequencing

# Quantified Jobshop


machines   
Papers 1, 2, 3

Table 1: Jobshop model notation.

$J$	set of jobs
$M$	set of machines
$T$	set of tasks, $T \subseteq J \times M$
$O$	taskorder, $O \subseteq T \times T$
$s_{j,m}$	start time (integer) of task $(j,m)$
$d_{j,m}$	duration of task $(j,m)$
$\delta_{j,m}$	additional duration of task $(j,m)$ in case of delay
$e_{j,m}$	earliness of task $(j,m)$ , i.e., $e = \max\{d^1 - d^2, 0\}$
$\bar{e}$	mean earliness
$m$	makespan
$r_{u,j,m}$	indicator of unary encoding of retarded task
$\tilde{r}_b$	indicator of binary encoding retarded task
$w$	wrapping indicator for binary to unary translation

Table 2: Jobshop tasks.

job	machine	duration	extra
Paper1	Blue	45	5
Paper1	Yellow	10	0
Paper2	Blue	20	5
Paper2	Green	10	10
Paper2	Yellow	34	0
Paper3	Blue	12	0
Paper3	Green	17	0
Paper3	Yellow	28	20

Table 3: Jobshop order.

prior task	later task
Paper1 Blue	Paper1 Yellow
Paper2 Green	Paper2 Blue
Paper2 Blue	Paper2 Yellow
Paper3 Yellow	Paper3 Blue
Paper3 Blue	Paper3 Green

# Quantified Jobshop



$$\begin{aligned}
 \min \quad & m^2 + k \cdot \bar{e} + \frac{1}{M} \cdot m^1 \quad \text{s.t.} \quad \exists s^1 y^1 m^1 \forall \tilde{r} \exists r w, s^2 y^2 m^2, e: \quad (1) \\
 & s_{j,m}^1 + d_{j,m} \leq m^1 \quad \forall (j,m) \in T \quad (2) \\
 & s_{i,m}^1 + d_{i,m} \leq s_{j,n}^1 \quad \forall (i,m,j,n) \in O \quad (3) \\
 & s_{i,m}^1 + d_{i,m} \leq s_{j,m}^1 + M \cdot (1 - y_{i,m,j}^1) \quad \forall (i,m) \in T, (j,m) \in T \quad (4) \\
 & s_{j,m}^1 + d_{j,m} \leq s_{i,m}^1 + M \cdot y_{i,m,j}^1 \quad \forall (i,m) \in T, (j,m) \in T \quad (5) \\
 & \sum_{(u,j,m) \in U} u \cdot r_{j,m} = \sum_{b \in B} 2^b \cdot \tilde{r}_b - |T| \cdot w \quad \wedge \quad \sum_{\substack{u \in U \\ (j,m) = T_u}} r_{j,m} \leq 1 \quad (6) \\
 & s_{j,m}^2 + d_{j,m} + \delta_{j,m} \cdot r_{j,m} \leq m^2 \quad \forall (j,m) \in T \quad (7) \\
 & s_{i,m}^2 + d_{i,m} + \delta_{i,m} \cdot r_{i,m} \leq s_{j,n}^2 \quad \forall (i,m,j,n) \in O \quad (8) \\
 & s_{i,m}^2 + d_{i,m} + \delta_{i,m} \cdot r_{i,m} \leq s_{j,m}^2 + M \cdot (1 - y_{i,m,j}^2) \quad \forall (i,m) \in T, (j,m) \in T \quad (9) \\
 & s_{j,m}^2 + d_{j,m} + \delta_{i,m} \cdot r_{i,m} \leq s_{i,m}^2 + M \cdot y_{i,m,j}^2 \quad \forall (i,m) \in T, (j,m) \in T \quad (10) \\
 & e_{i,m} \geq s_{i,m}^1 - s_{i,m}^2 \quad \forall (i,m) \in T \quad (11) \\
 & e_{i,m} \geq 0 \quad (12) \\
 & \bar{e} = \frac{1}{|T|} \cdot \sum_{(i,m) \in T} e_{i,m} \quad (13)
 \end{aligned}$$

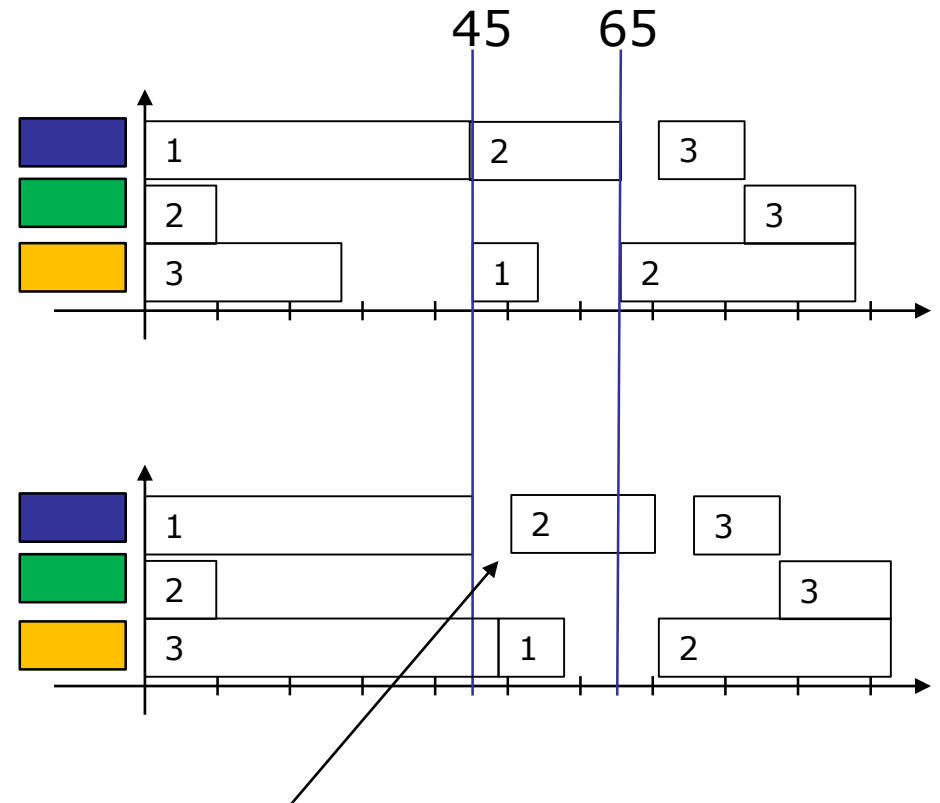
bin. encoding  
of scenario-index

4  
↓  
01000

# Quantified Jobshop

Table 4: Solution of the Jobshop Example.

sc.	start times								m.s.
	1B	1Y	2G	2B	2Y	3Y	3B	3G	
	<i>first stage solution</i>								
	0	45	0	45	65	0	70	82	99
1	0	50	0	50	70	0	70	82	104
2	0	45	0	45	65	0	65	82	99
3	0	45	0	45	70	0	70	82	104
4	0	45	0	45	65	0	65	82	99
5	0	45	0	45	65	0	65	82	99
6	0	45	0	45	65	0	65	82	99
7	0	45	0	45	65	0	70	82	99
8	0	48	0	50	70	0	75	87	104



A degree of freedom, i.e. this is no worst-case scenario.



# Quantified Jobshop

Optimal first-stage solution

0 45 0 45 65 0 70 82 --- 99

Scenario 8:

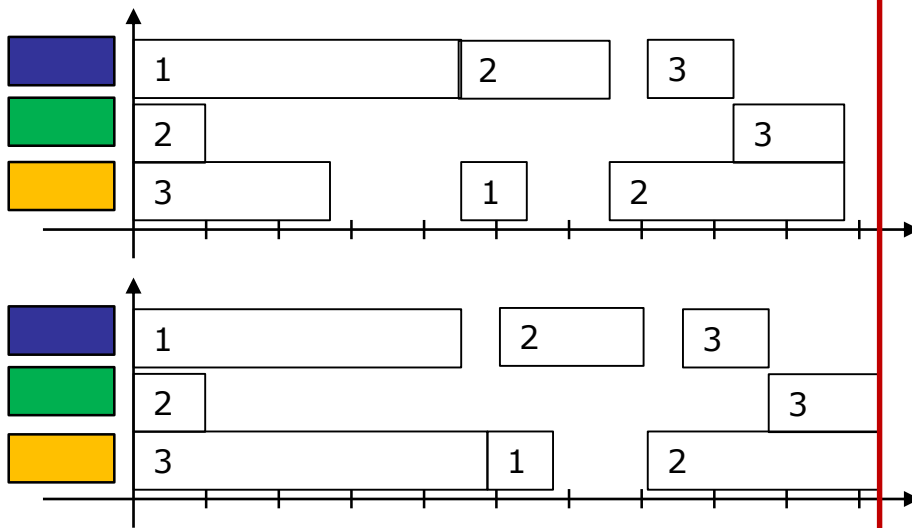
0 48 0 50 70 0 75 87 --- 104

Optimal deterministic solution

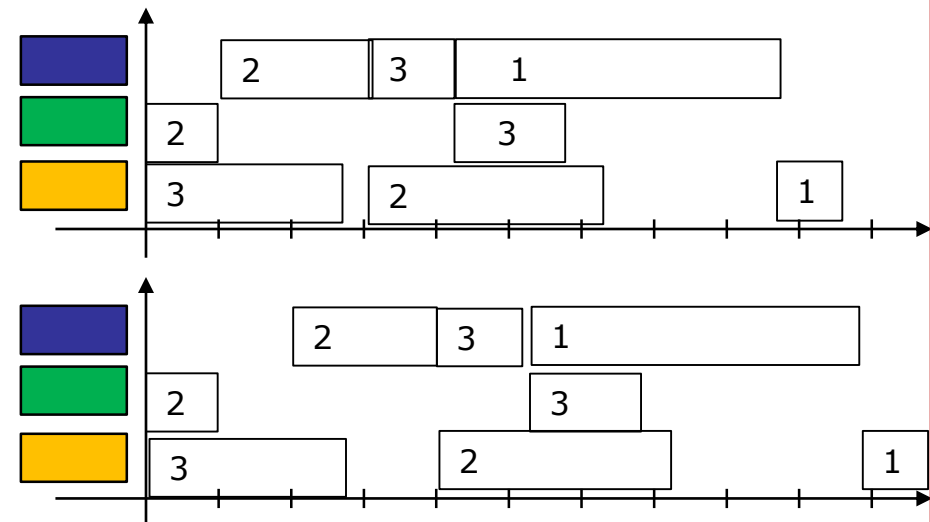
42 87 0 10 30 0 30 42 --- 97

Scenario 4:

52 97 0 20 40 0 40 52 --- 108



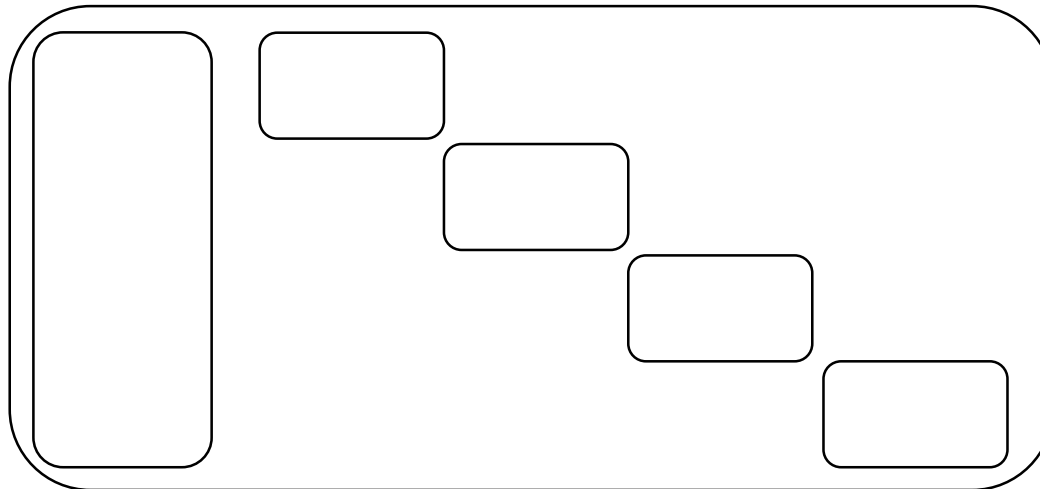
No scenario is worse than: 104



Without disruption awareness: 108

## Opportunity 1: Deterministic equivalent

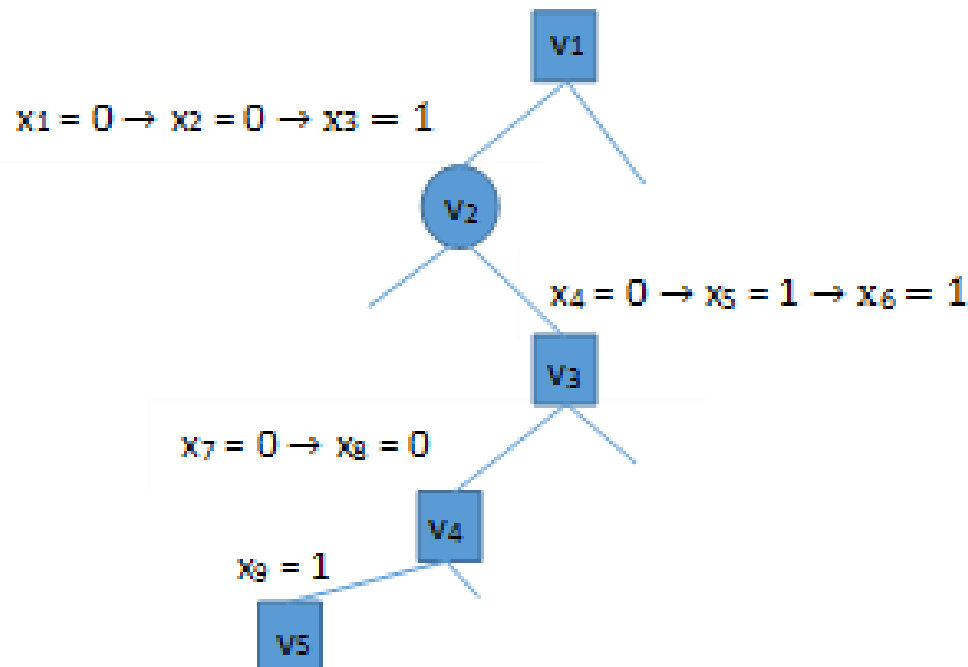
fst-stage | scenarion variables



- >> n universal variables
- >> n „uncertainty bits“
- >>  $2^n$  many scenarios

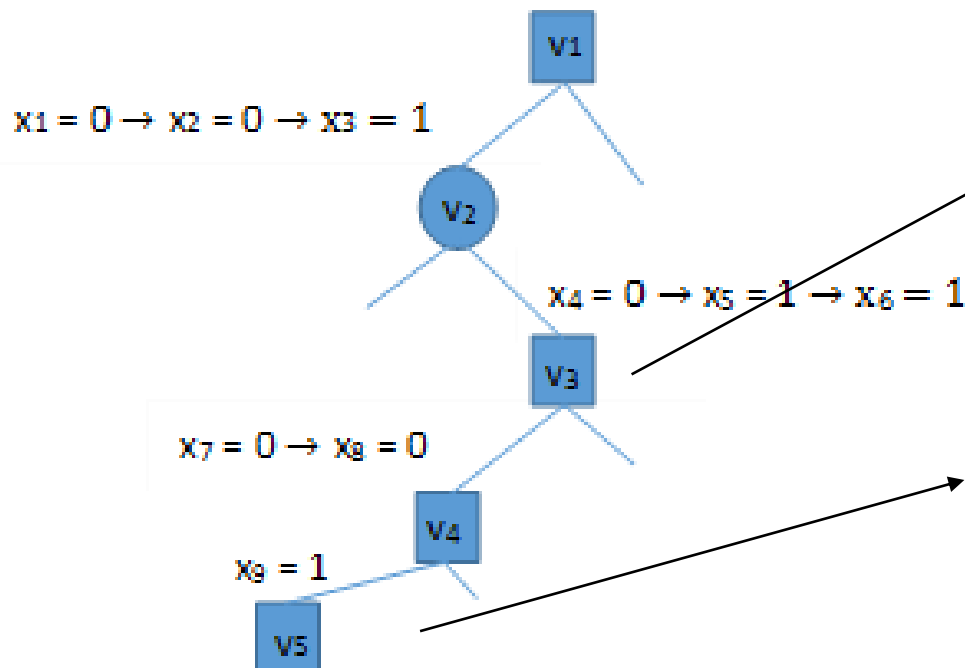
## Opportunity 2: Tree search and cut algorithm

starting with 0/1-QIPs, run into tree from root downwards



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starting with 0/1-QIPs, run into tree from root downwards

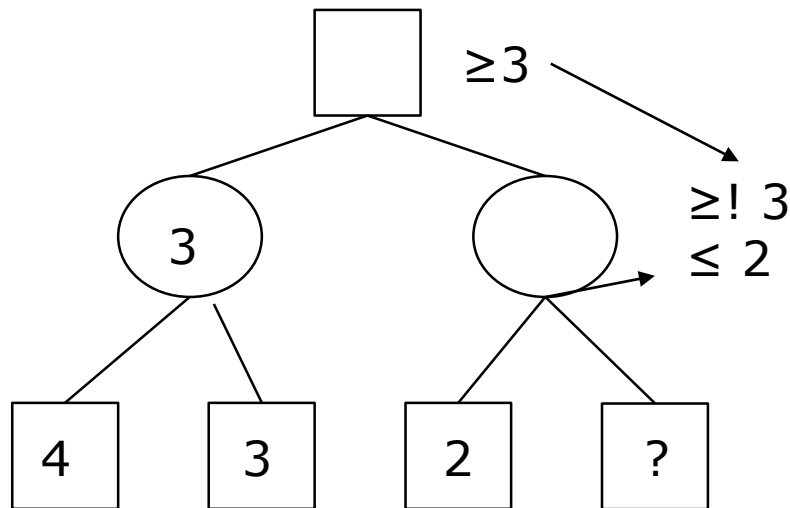


solve LP-relaxation  
if feasible:  
strengthen with cuts  
if infeasible:  
compute backjump

if contradiction detected:  
compute backjump and  
learn cover-cut  
discover earlier implication

## Opportunity 2: Tree search and cut algorithm

starting with 0/1-QIPs, run into tree from root downwards



Best first search algorithm?

Currently depth first search  
-> Alpha algorithm

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## Algorithm 1: A basic alphabeta(int d, int a, int b) routine, sketched

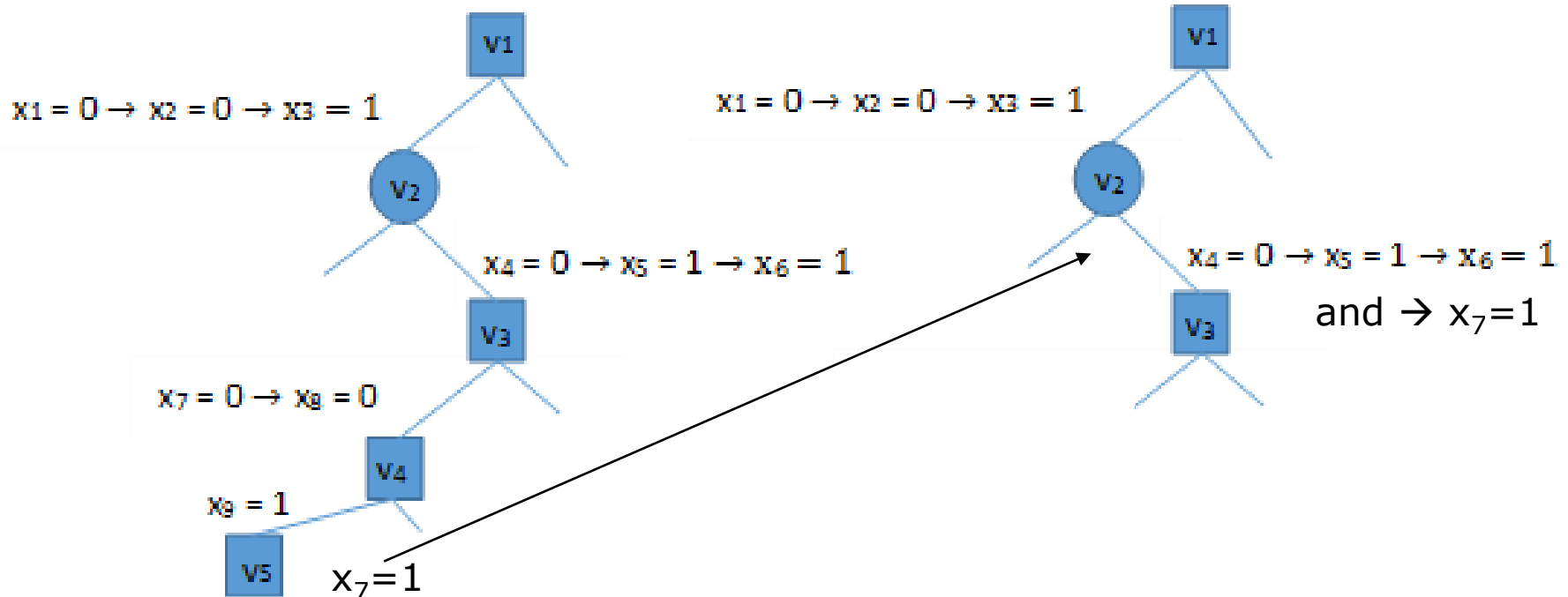
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```
1 compute LP-relaxation, solve it, extract branching variable or cut;
2 if integer solution found then return objective value ; // leaf reached
3 if  $x_i$  is an existential variable then score :=  $-\infty$ ; else score :=  $+\infty$ ;
4 for val_ix from 0 to 1 do // search actually begins ...
5     if level_finished(t) then // leave marked recursion levels
6         if  $x_i$  is an existential variable then return score ;
7         else return  $-\infty$  ;
8     end
9     assign( $x_i$ , val[val_ix], ... );
10    v := alphabeta(d-1, fmax(a, score), b);
11    unassign( $x_i$ );
12    if  $x_i$  is an existential variable then
13        if  $v > score$  then score := v; // existential player maximizes
14        if score  $\geq b$  then return score ;
15    else
16        ... analogously ...;
17    end
18 end
```

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## Opportunity 2: Tree search and cut algorithm

starting with 0/1-QIPs, run into tree from root downwards



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**Algorithm 2:** A local repetition loop extends the basic algorithm with conflict analysis and learning; replaces line 9 in Algorithm 1

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```
1 repeat
2   if the current level is marked as finished then leave the recursion level;
3   if propagate(confl, confl_var, confl_partner, false) // unit prop. [11] then
4     if  $x_i$  is an existential variable then
5       v = alpha(t+1, lsd-1, fmax(a, score), b);
6       if v > score then score := v;
7       if score  $\geq$  b then break;
8     else
9       ... analogously ...;
10    end
11  else
12    add reason, i.e. a constraint, to the database ; // [11]
13    returnUntil(out_target_dec_level) ; // set level_finished(...)
14    if  $x_i$  is an existential variable then return score; else return  $-\infty$  ;
15  end
16 until there is no backward implied variable;
```



# Yasol results

# univ. var.		Depqbf	DEPCplex	Yasol
1-5 UV	Time	39185	41133s	73732
	#solved	312/373	315/373	259/373
6-10 UV	Time	1805s	1351s	2146s
	#solved	38/41	39/41	38/41
11-15 UV	Time	7961s	25297s	11023s
	#solved	84/97	65/97	79/97
16-20 UV	Time	2609s	84685s	5929s
	#solved	166/170	37/170	167/170
21+ UV	Time	16856s	69600s <sup>3</sup>	39620s
	#solved	96/116	0/116	60/116
$\Sigma$	Time	68416s	194128s	132350s
$\Sigma$	#solved	696/797	456/797	603/797

A collection of 797 QBF instances, grouped concerning the instances' number of universal variables.

# Yasol results

No. UV	DEPCplex	Yasol	DEPScip	DEPCbc
0 (max 1h)	59 / 59 19520s	24 / 59 132129s	48 / 59 55062s	34 / 59 112958s
1 to 4 (max 900s)	138 / 138 42s	103 / 138 39984s	137 / 138 1072s	138 / 138 354s
10 to 14 (max 1200s)	46 / 46 1854s	46 / 46 972s	34 / 46 23440s	23 / 46 38867s

A collection of 0/1-IPs and artificially constructed 0/1-QIPs, grouped concerning the instances' number of universal variables.

- existential integer variables
- continuous variables in the last stage
- more sophisticated cutting planes
- more models
- best-first search

THANKS FOR YOUR ATTENTION