

Planning via **Model Checking**



AALBORG UNIVERSITET

“Plan”

- Timed Automata
 - Model Checking
 - Time-optimal Planning
 - Zones
- Priced Timed Automata
 - Cost-optimal Planning
 - Priced Zones A*
- Timed Games
 - Dynamic Planning
 - Strategies, Zones
- Stochastic Timed Automata
 - Performance Analysis
 - Statistical Model Checking
- Stochastic Priced Timed Games
 - Expected Cost Optimal Adaptive Planning
 - Strategies
 - Reinforcement Learning
- Applications



Verification

CLASSIC

1995

Optimization

CORA

2001

Synthesis

TIGA

2005

Component

ECDAR

2010

Testing

TRON

2004

Performance
Analysis

SMC

2011

2014

STRATEGO



Priced Timed Automata

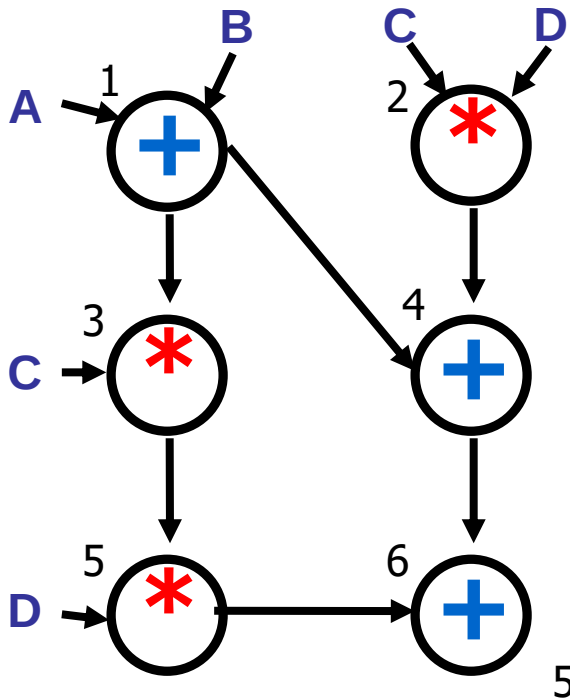
Cost-Optimal Planning



AALBORG UNIVERSITET



Task Graph Scheduling – Revisited



Compute :
 $(D * (C * (A + B)) + ((A + B) + (C * D)))$

using 2 processors

P1 (fast)



ENERGY:

10

P2 (slow)



20

+	2ps
*	3ps

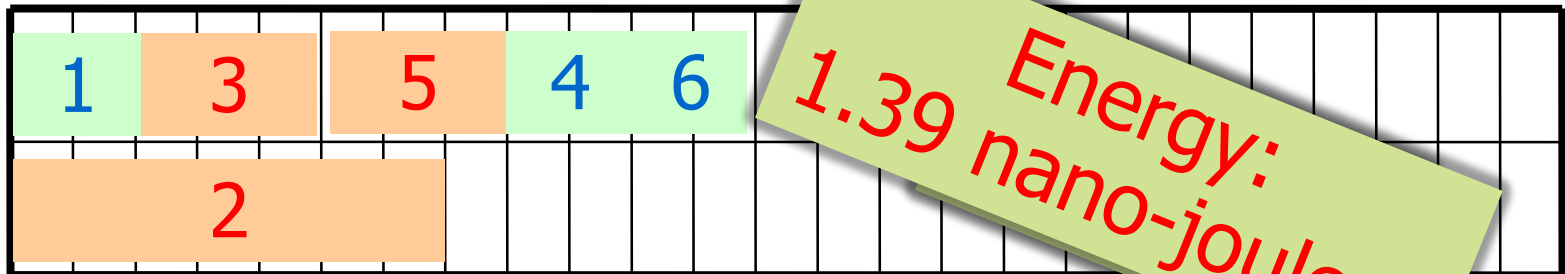
Idle	10W
In use	90W

+	5ps
*	7ps

Idle	20W
In use	30W

P1

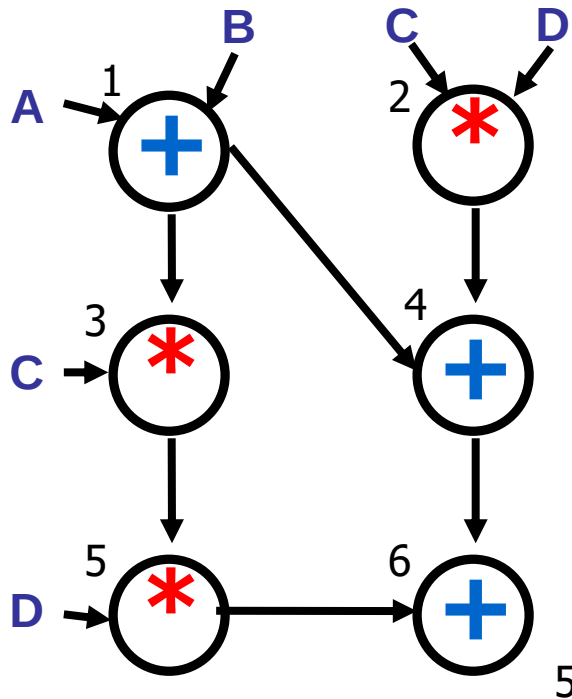
P2



Energy: 1.39 nano-joule !!



Task Graph Scheduling – Revisited



Compute :
 $(D * (C * (A + B)) + ((A + B) + (C * D)))$

using 2 processors

P1 (fast)



ENERGY:
10

+	2ps
*	3ps

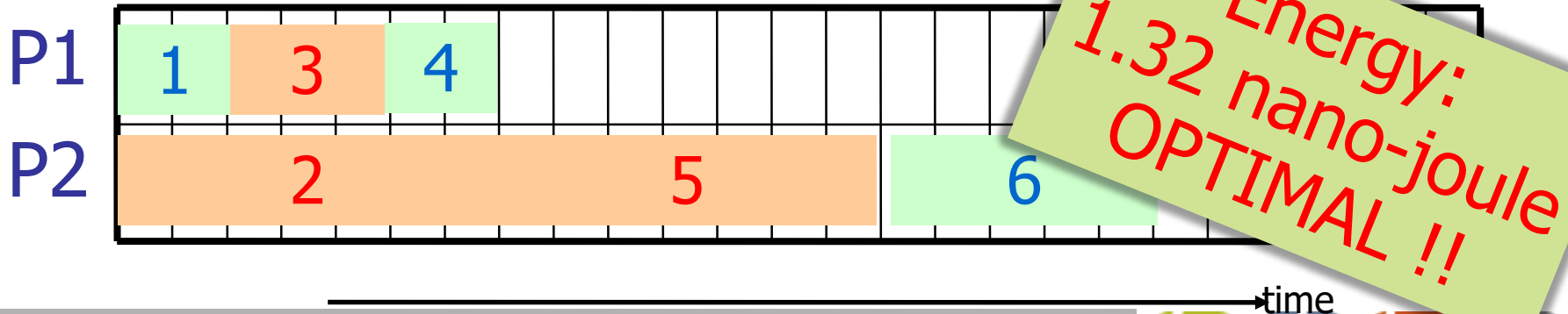
Idle	10W
In use	90W

P2 (slow)



+	5ps
*	7ps

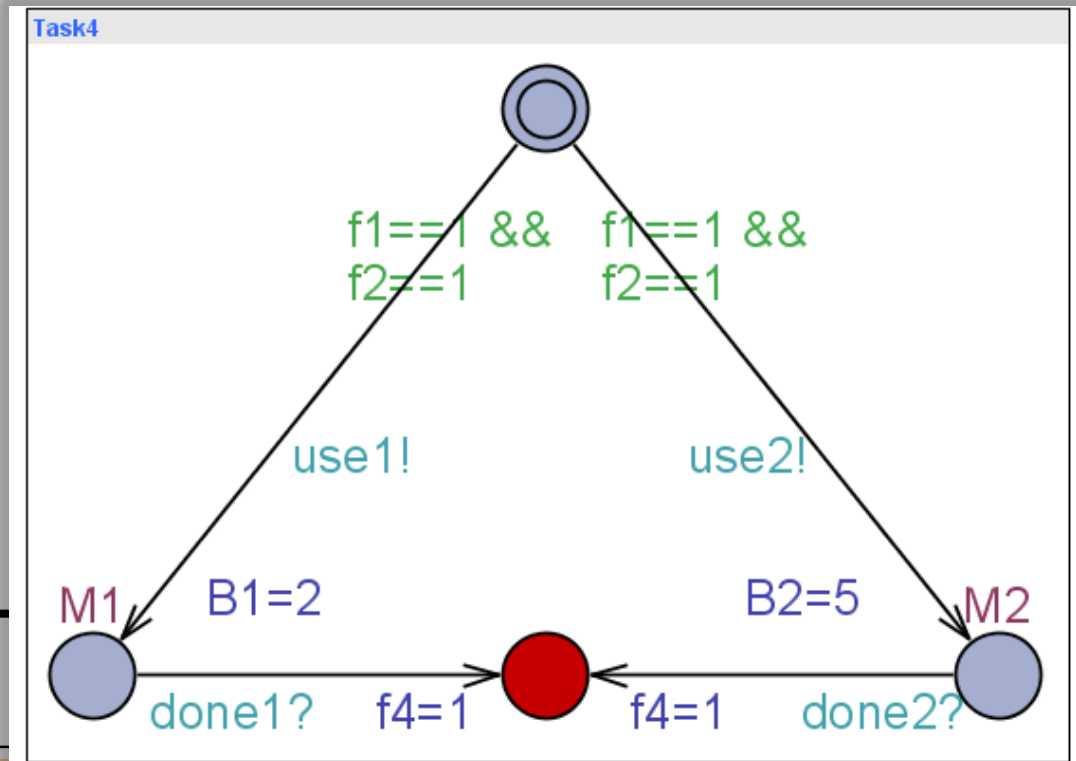
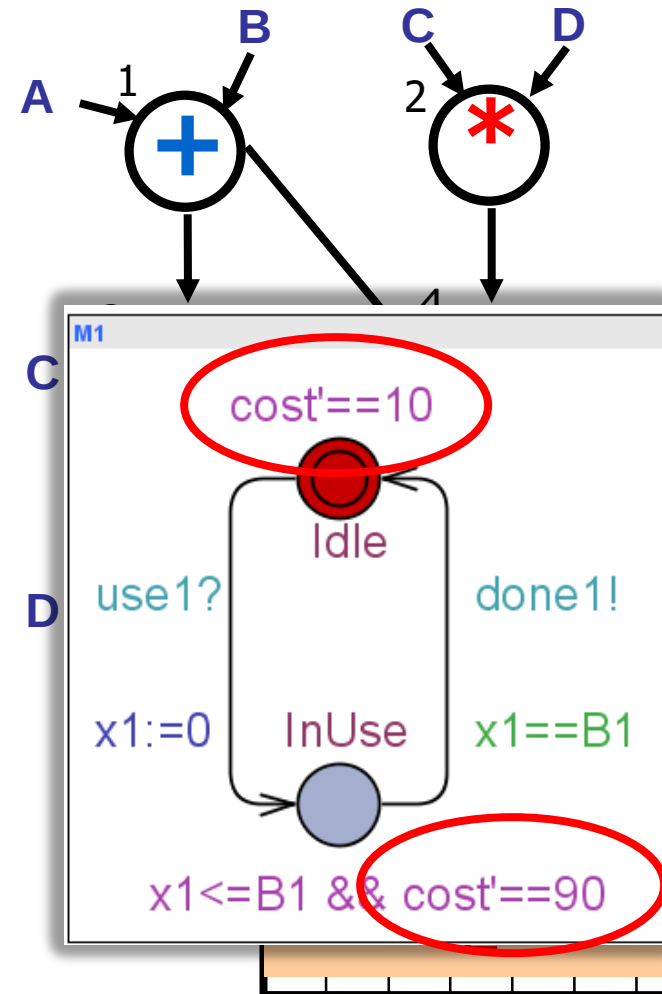
Idle	20W
In use	30W



Task Graph Scheduling – Revisited

Compute :

$$(D * (C * (A + B)) + ((A + B) + (C * D)))$$



5

6

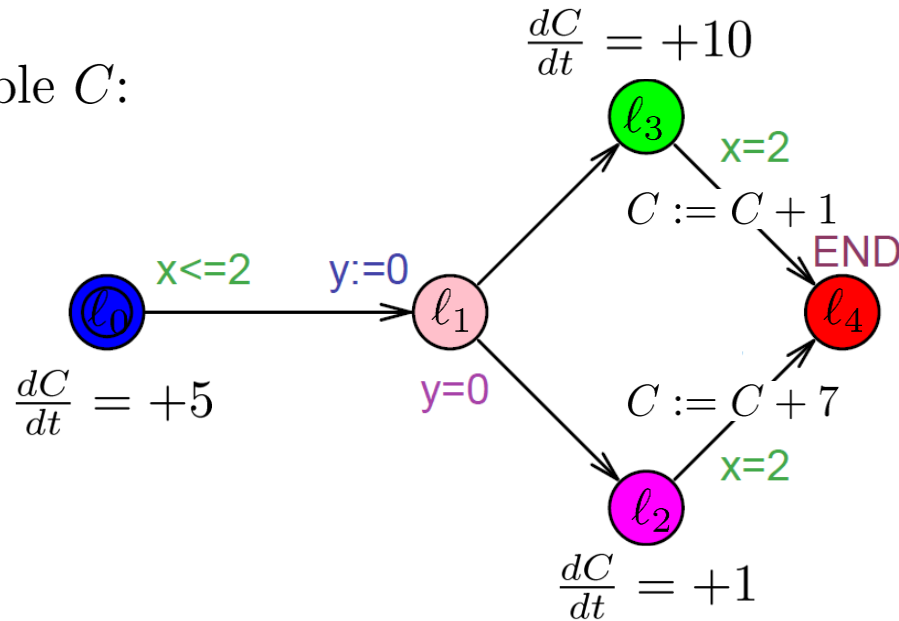
OPTIMAL !!

time



A simple example

Observer variable C :



$$(\ell_0, [0, 0]) \xrightarrow{1.9}_{9.5} (\ell_0, [1.9, 1.9]) \rightarrow_0 (\ell_1, [1.9, 0]) \rightarrow_0$$

$$(\ell_2, [1.9, 0]) \xrightarrow{0.1}_{0.1} (\ell_2, [2, 0.1]) \rightarrow_7 (\ell_4, [2, 0.1])$$

$$\sum C_i = 16.6$$

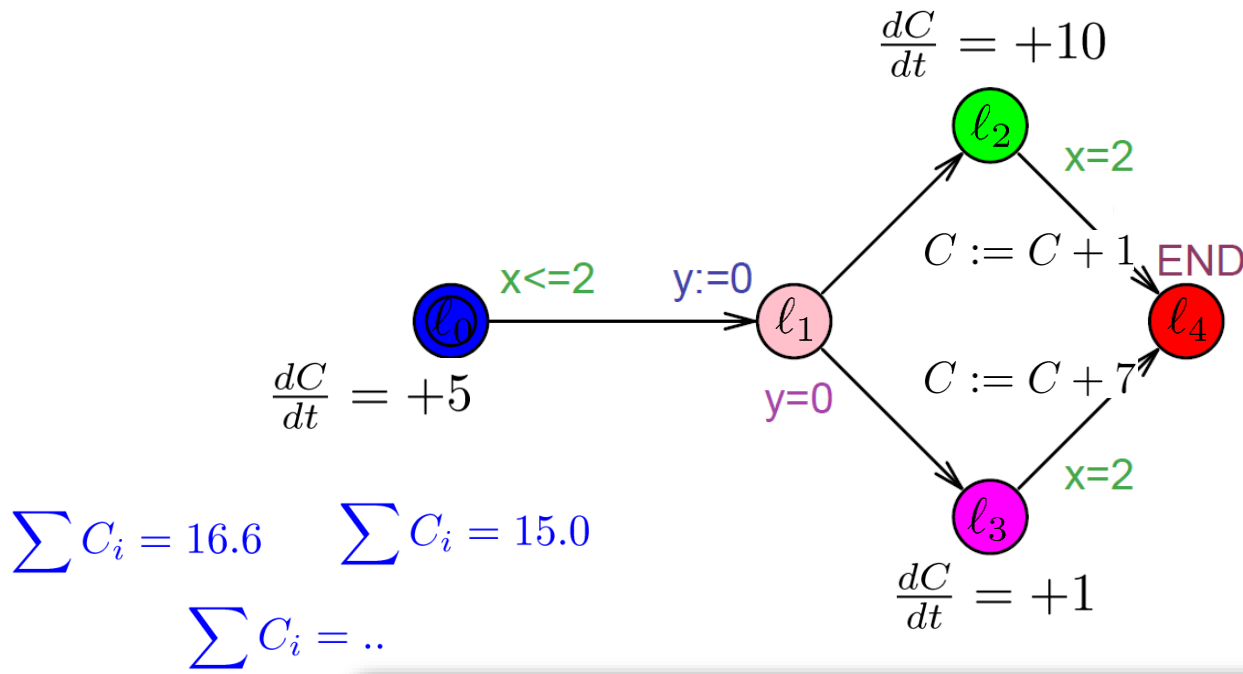
$$(\ell_0, [0, 0]) \xrightarrow{1.2}_{6.0} (\ell_0, [1.2, 1.2]) \rightarrow_0 (\ell_1, [1.2, 0]) \rightarrow_0$$

$$(\ell_3, [1.2, 0]) \xrightarrow{0.8}_{8.0} (\ell_3, [2, 0.8]) \rightarrow_1 (\ell_4, [2, 0.8])$$

$$\sum C_i = 15.0$$



A simple example

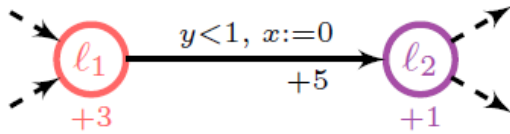


Q: What is cheapest cost for reaching l_4 ?

$$\inf_{0 \leq t \leq 2} \min\{5t + 10(2 - t) + 1, 5t + (2 - t) + 4\} = 9$$

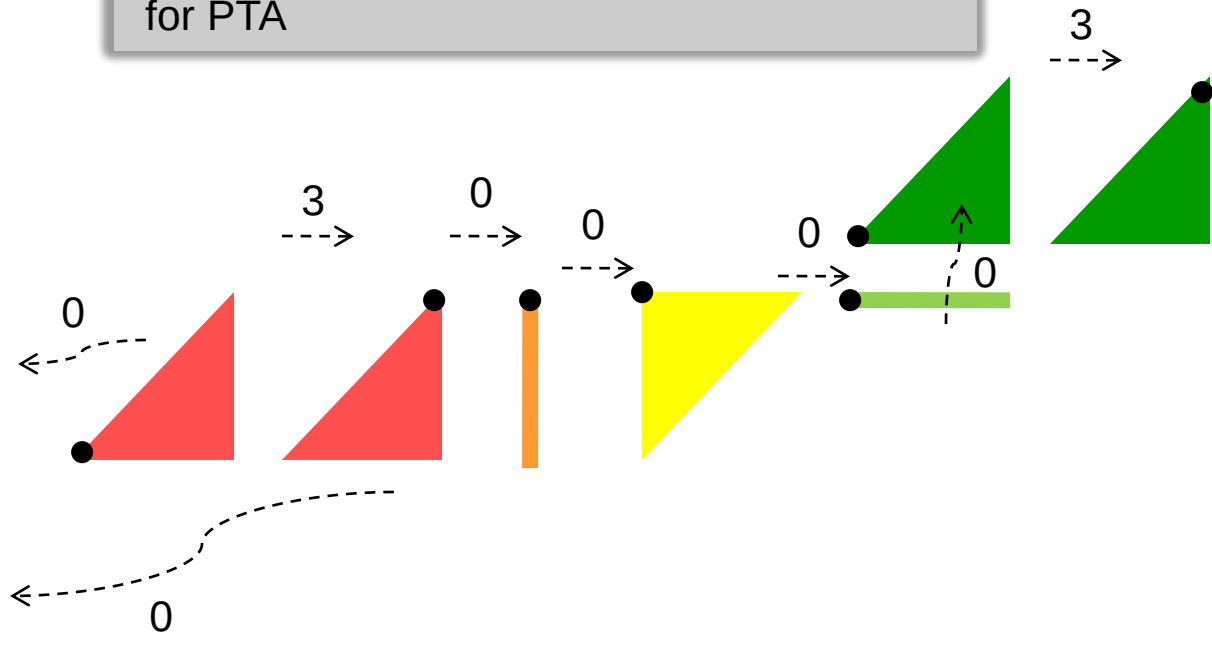
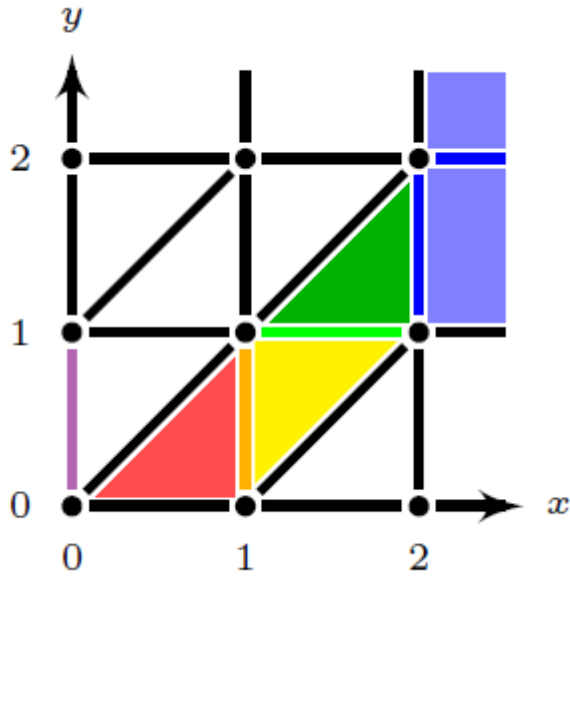
→ **strategy:** leave immediately l_0 , go to l_3 , and wait there 2 t.u.

Corner Point Regions



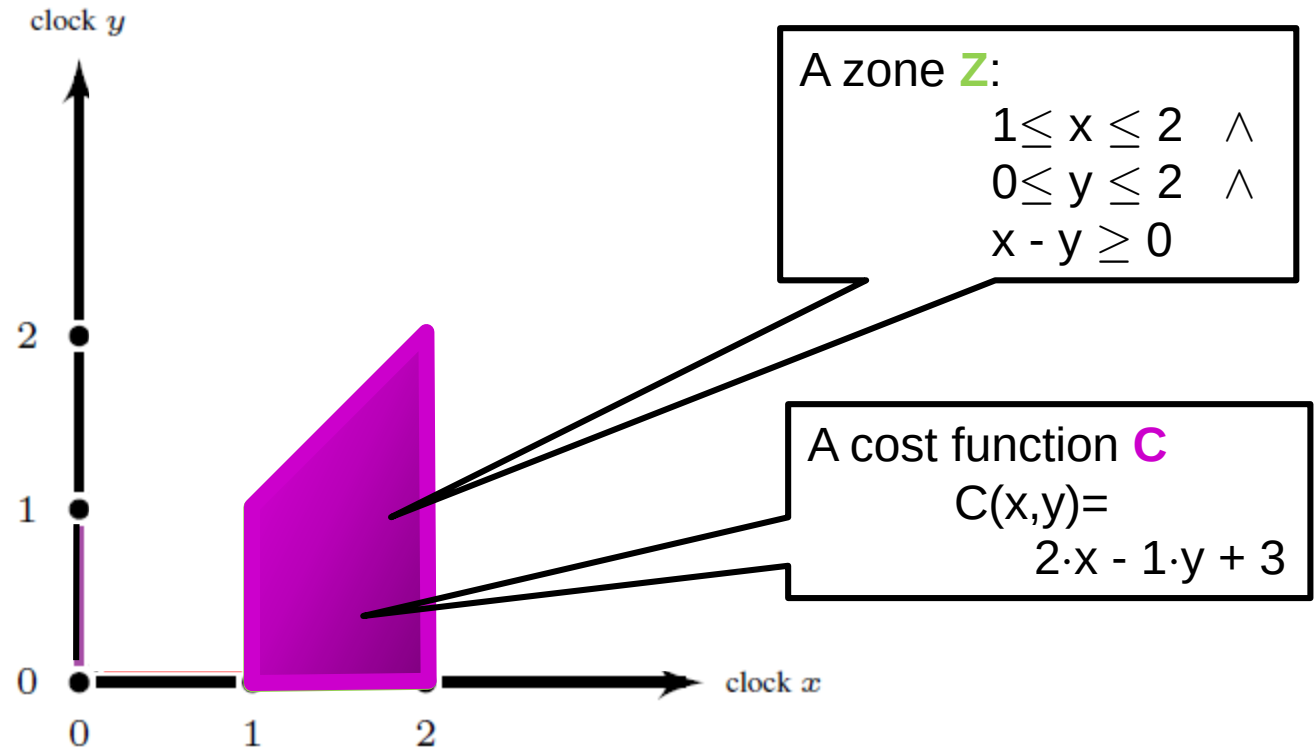
THM [Behrmann, Fehnker ..01] [Alur,Torre,Pappas 01]
Optimal reachability is decidable for PTA

THM [Bouyer, Brojaue, Briuere, Raskin 07]
Optimal reachability is PSPACE-complete
for PTA



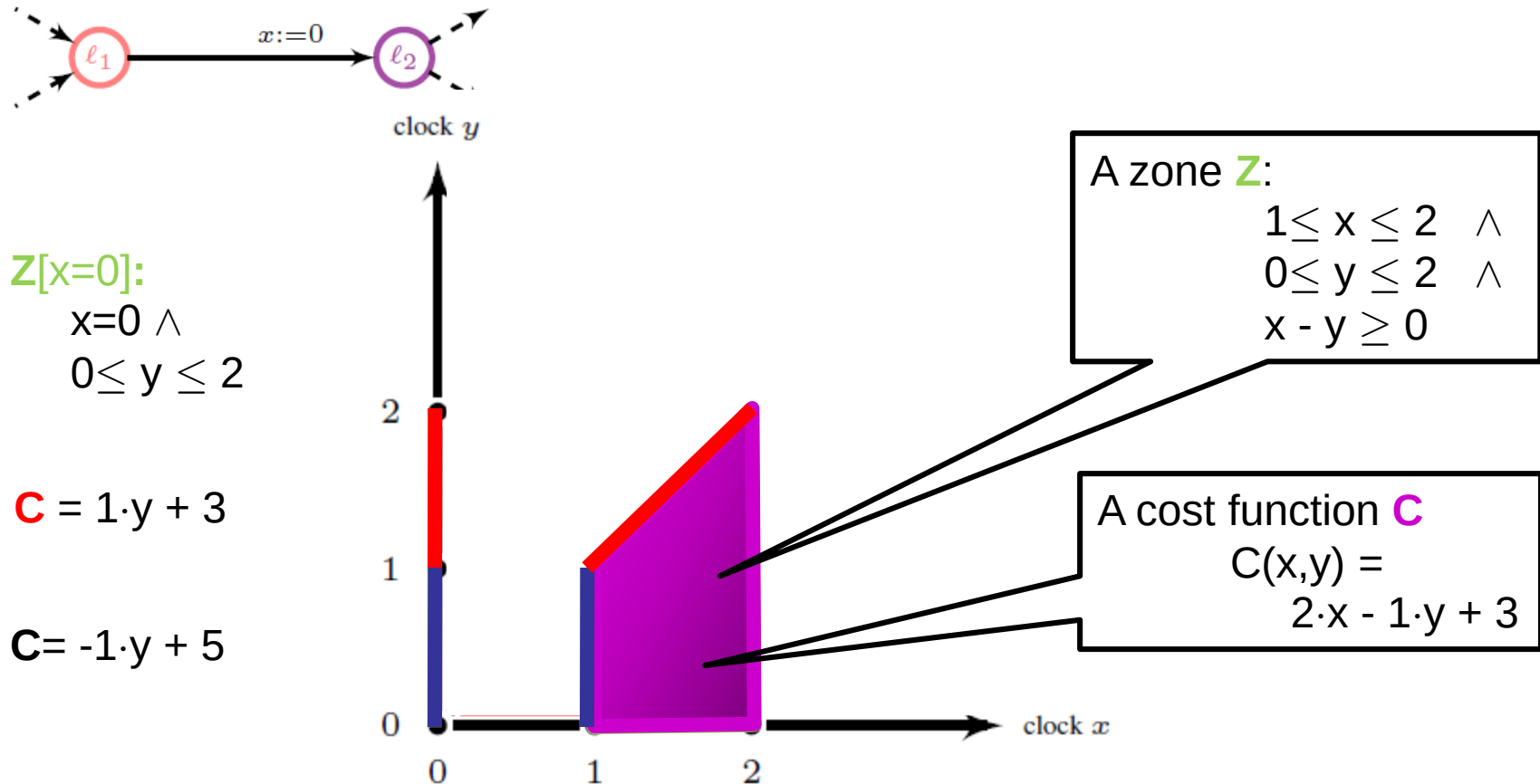
Priced Zones

[CAV01]



Priced Zones – Reset

[CAV01]



Symbolic Branch & Bound Algorithm

Cost := ∞

Passed := $\emptyset$

Waiting := $\{(l_0, Z_0)\}$

while Waiting $\neq \emptyset$ **do**

select (l, Z) from Waiting

if $l = l_g$ and $\text{minCost}(Z) < \text{Cost}$ **then**

 Cost := $\text{minCost}(Z)$

if $\text{minCost}(Z) + \text{Rem}_{(l,Z)} \geq \text{Cost}$ **then**

if for all (l, Z') in Passed: $Z' \not\leq Z$ **then**

add (l, Z) to Passed

add all (l', Z') with $(l, Z) \rightarrow (l', Z')$

return Cost

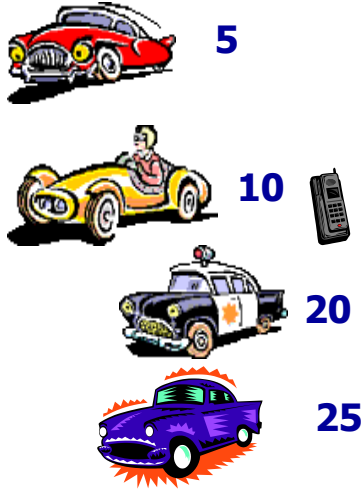
$$Z' \leq Z$$

Z' is bigger & cheaper than Z

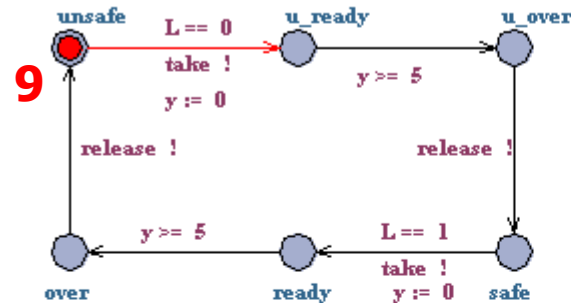
$\leq$ is a well-quasi ordering which guarantees termination!



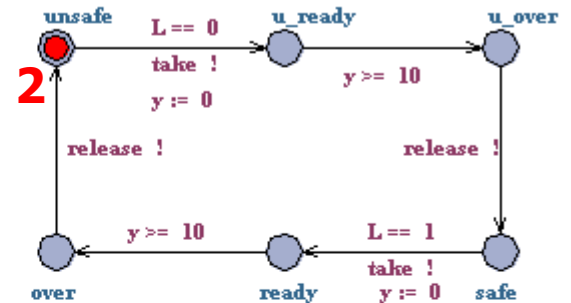
EXAMPLE: Optimal rescue plan for cars with different subscription rates for city driving !



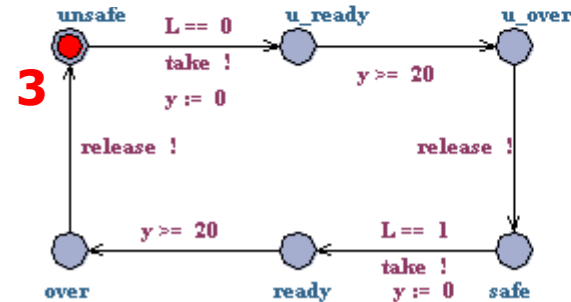
Golf



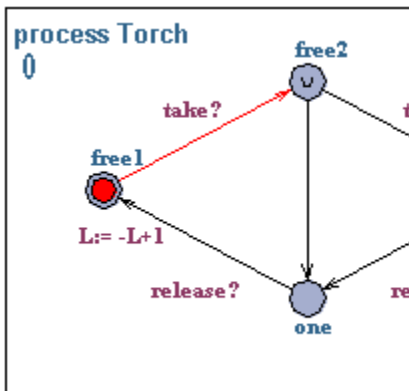
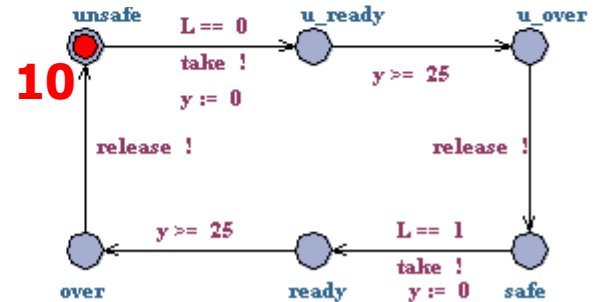
Citroen



BMW



Datsun



OPTIMAL PLAN HAS ACCUMULATED **COST**=195 and **TOTAL TIME**=65!

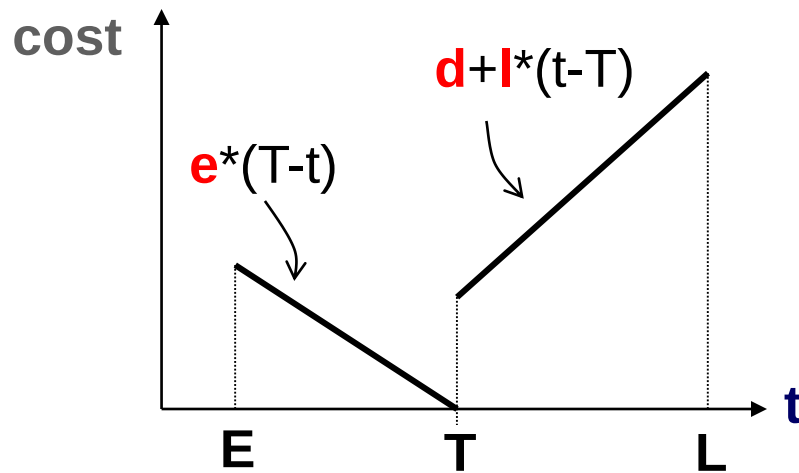


Experiments

COST-rates				SCHEDULE	COST	TIME	#Expl	#Pop'd
G	C	B	D					
Min Time				CG> G< BD> C< CG>		60	1762 1538	2638
1	1	1	1	CG> G< BG> G< GD>	55	65	252	378
9	2	3	10	GD> G< CG> G< BG>	195	65	149	233
1	2	3	4	CG> G< BD> C< CG>	140	60	232	350
1	2	3	10	CD> C< CB> C< CG>	170	65	263	408
1	20	30	40	BD> B< CB> C< CG>	975 1085	85 time<85	-	-
0	0	0	0	-	0	-	406	447

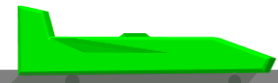


Example: Aircraft Landing



- E** earliest landing time
- T** target time
- L** latest time
- e** cost rate for being early
- l** cost rate for being late
- d** fixed cost for being late

Planes have to keep separation distance to avoid turbulences caused by preceding planes



$$\text{minimize } \sum_{i=1}^P (e_i \alpha_i + d_i \beta_i)$$

where

t_i : landing time of plane i

α_i : how early plane i lands before target T_i

β_i : how late plane i lands after target T_i

δ_{ij} : if i lands before j then 0 otherwise 1



Modelling ALP with MILP

[Beasley00]

$E_i \leq t_i \leq L_i$: Land only in the permitted slot

$\delta_{ij} + \delta_{ji} = 1$: Either or

$t_i \geq t_j + S_{ij} - M\delta_{ij}$: Respect the turbulence

$\alpha_i \geq T_i - t_i$: Bigger than the difference between T_i and t_i

$0 \leq \alpha_i \leq T_i - E_i$: The relevant interval

$\beta_i \geq t_i - T_i$: Bigger than the difference between t_i and T_i

$0 \leq \beta_i \leq L_i - T_i$: The relevant intervals

$t_i = T_i - \alpha_i + \beta_i$: Guaranteeing the right relationships

• Total formulation

- $3P$ continuous variables
 - $2P(P - 1) + PR$ binary variables
 - $4P + (4 + R)P(P - 1)/2$ constraints
- 2 runways, 30 planes $\Rightarrow$ 1890 variables 2730 constraints

• Additional tightening constraints to help LP relaxation

Automated Planning Tools, ...

minimize $\sum_{i=1}^P (e_i \alpha_i + d_i \beta_i)$

where

t_i : landing time of plane i

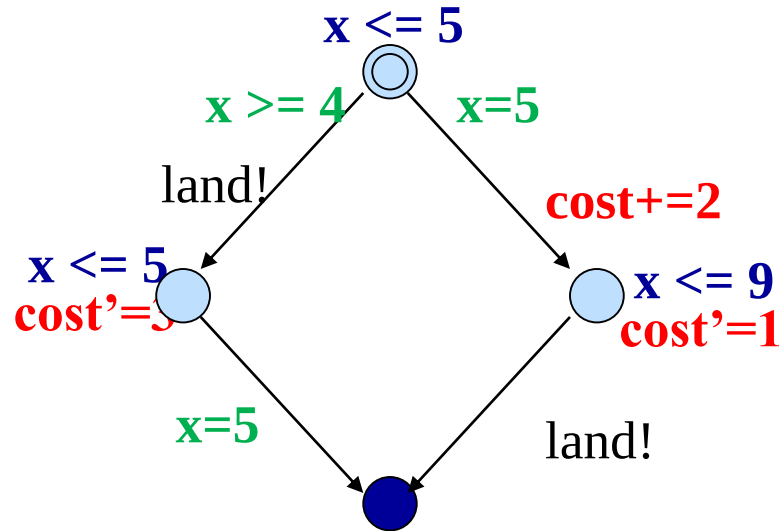
α_i : how early plane i lands before target T_i

β_i : how late plane i lands after target T_i

δ_{ij} : if i lands before then 0 otherwise 1



Aircraft Landing – PTA



- 4** earliest landing time
- 5** target time
- 9** latest time
- 3** cost rate for being early
- 1** cost rate for being late
- 2** fixed cost for being late

Planes have to keep separation distance to avoid turbulences caused by preceding planes

Aircraft Landing

Source of examples:
Baesley et al'2000

	problem instance	1	2	3	4	5	6	7
	number of planes	10	15	20	20	20	30	44
	number of types	2	2	2	2	2	4	2
1	optimal value	700	1480	820	2520	3100	24442	1550
	explored states	481	2149	920	5693	15069	122	662
	cputime (secs)	4.19	25.30	11.05	87.67	220.22	0.60	4.27
2	optimal value	90	210	60	640	650	554	0
	explored states	1218	1797	669	28821	47993	9035	92
	cputime (secs)	17.87	39.92	11.02	755.84	1085.08	123.72	1.06
3	optimal value	0	0	0	130	170	0	
	explored states	24	46	84	207715	189602	62	N/A
	cputime (secs)	0.36	0.70	1.71	14786.19	12461.47	0.68	
4	optimal value				0	0		
	explored states	N/A	N/A	N/A	65	64	N/A	N/A
	cputime (secs)				1.97	1.53		



Symbolic Branch & Bound Algorithm

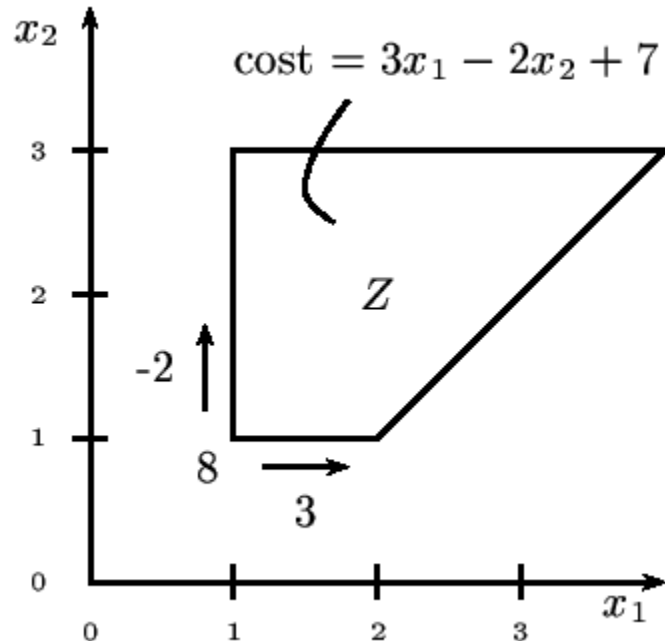
```
Cost :=  $\infty$ 
Passed :=  $\emptyset$ 
Waiting :=  $\{(l_0, Z_0)\}$ 
while Waiting  $\neq \emptyset$  do
    select  $(l, Z)$  from Waiting
    if  $l = l_g$  and  $\text{minCost}(Z) < \text{Cost}$  then
        Cost :=  $\text{minCost}(Z)$ 
    if  $\text{minCost}(Z) + \text{Rem}_{(l, Z)} \geq \text{Cost}$  then break
    if for all  $(l', Z')$  in Passed:  $Z' \not\subseteq Z$  then
        add  $(l, Z)$  to Passed
        add all  $(l', Z')$  with  $(l, Z) \rightarrow (l', Z')$  to Waiting
return Cost
```

Zone based
Linear Programming
Problems
→(dualize)
Min Cost Flow

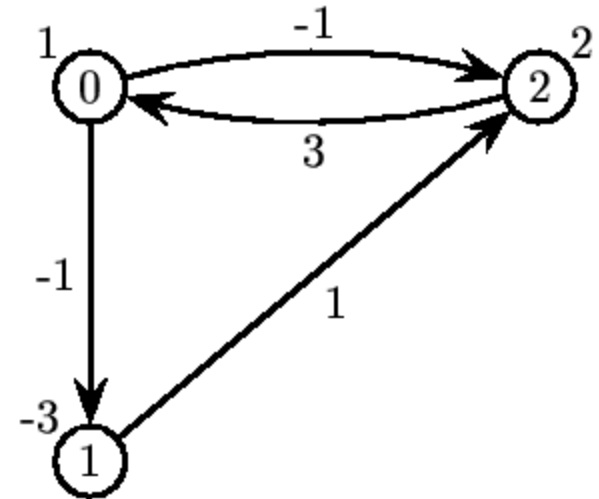


Zone LP $\rightarrow$ Min Cost Flow

Exploiting duality



minimize $3x_1 - 2x_2 + 7$
 when $x_1 - x_2 \leq 1$
 $1 \leq x_2 \leq 3$
 $x_2 \geq 1$

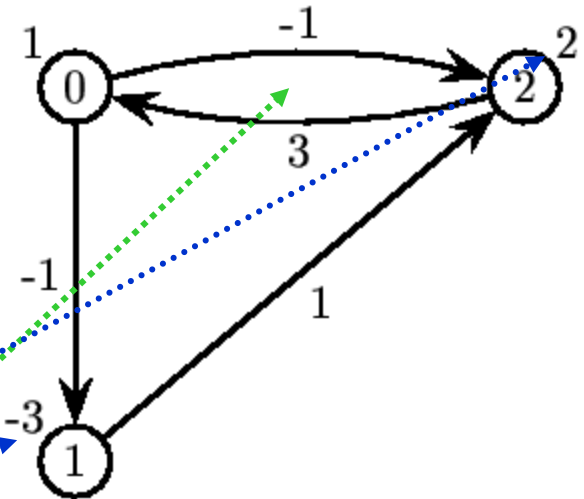
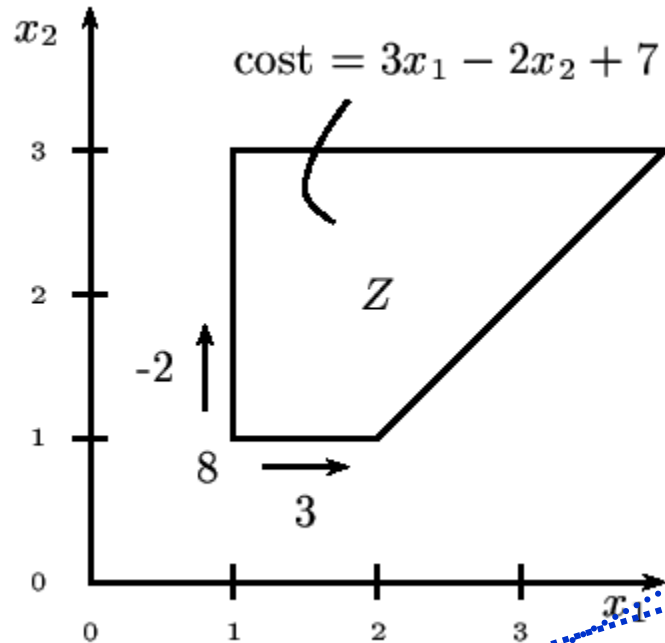


minimize $3y_{2,0} - y_{0,2} + y_{1,2} - y_{0,1}$
 when $y_{2,0} - y_{0,1} - y_{0,2} = 1$
 $y_{0,2} + y_{1,2} = 2$
 $y_{0,1} - y_{1,2} = -3$



Zone LP $\rightarrow$ Min Cost Flow

Exploiting duality



minimize $3x_1 - 2x_2 + 7$
 when $x_1 - x_2 \leq 1$
 $1 \leq x_2 \leq 3$
 $x_2 \geq 1$

minimize $3y_{2,0} - y_{0,2} + y_{1,2} - y_{0,1}$
 when $y_{2,0} - y_{0,1} - y_{0,2} = 1$
 $y_{0,2} + y_{1,2} = 2$
 $y_{0,1} - y_{1,2} = -3$



Aircraft Landing (revisited)

[TACAS04]

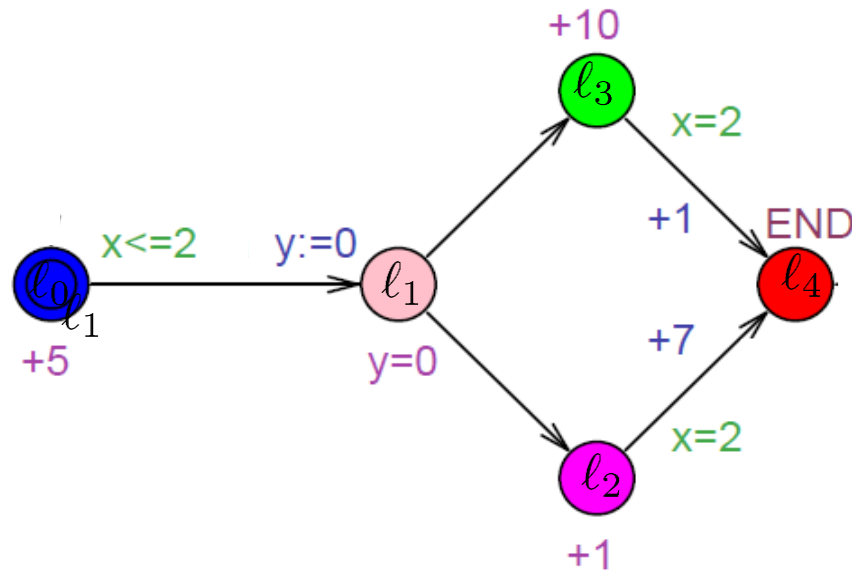
RW	Planes	10	15	20	20	20	30	44
	Types	2	2	2	2	2	4	2
1	simplex	0.844s	5.210s	2.135s	17.888s	44.878s	0.451s	0.670s
	netsimplex	0.156s	0.657s	0.369s	2.363s	5.503s	0.127s	0.322s
factor		5.41	7.93	5.79	7.57	8.16	3.55	2.08
2	simplex	2.577s	7.436s	2.175s	94.357s	120.004s	2.322s	0.264s
	netsimplex	0.332s	1.036s	0.436s	13.376s	18.033s	0.600s	0.179s
factor		8.00	7.18	4.99	7.054	6.65	3.87	1.474
3	simplex	0.120s	0.181s	0.357s	740.100s	516.678s	0.166s	N/A
	netsimplex	0.064s	0.104s	0.129s	170.176s	124.805s	0.079s	N/A
factor		1.87	1.74	2.77	4.34	4.14	2.10	
4	simplex	N/A	N/A	N/A	1.603s	0.318s	N/A	N/A
	netsimplex	N/A	N/A	N/A	0.378s	0.093s	N/A	N/A
factor					4.24	3.42		

A. Loebel (2000). MCF Version 1.2 - A network simplex implementation. (<http://www.zib.de>)



Optimal

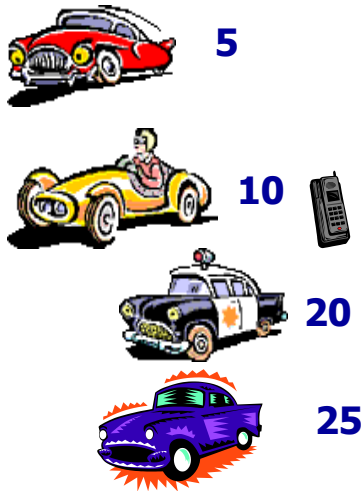
Schedule



$(\ell_0, [0, 0]) \xrightarrow{1.2}_{6.0} (\ell_0, [1.2, 1.2]) \rightarrow_0 (\ell_1, [1.2, 0]) \rightarrow_0$
 $(\ell_3, [1.2, 0]) \xrightarrow{0.8}_{8.0} (\ell_3, [2, 0.8]) \rightarrow_1 (\ell_4, [2, 0.8])$
 $\rightarrow_{2.0} (\ell_0, [0, 0])$

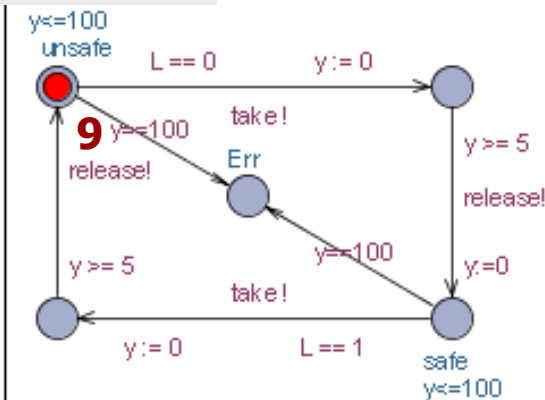
$$\sum_i C_i / \sum_i t_i = 17/2 = 8.5$$

EXAMPLE: Optimal WORK plan for cars with different subscription rates for city driving !

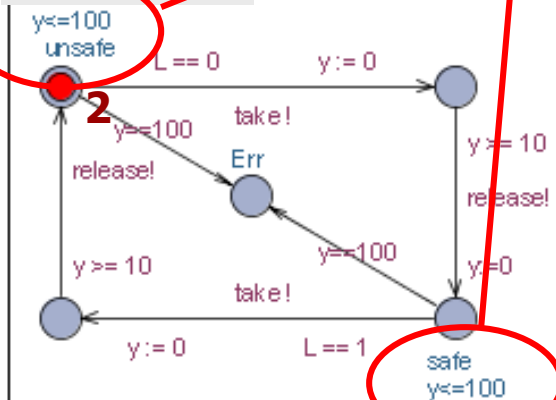


maximal 100 min.
at each location

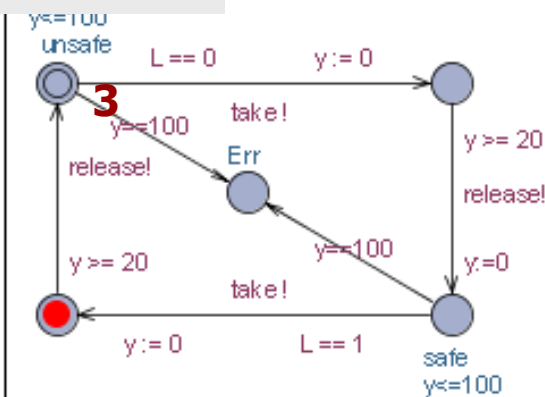
Golf



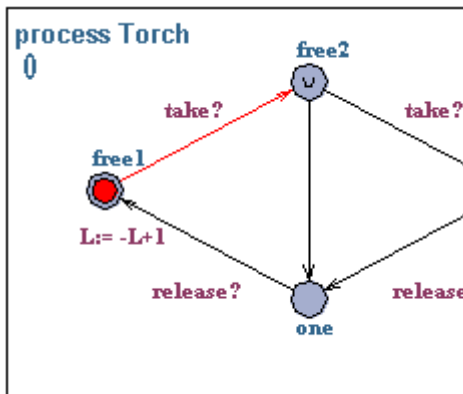
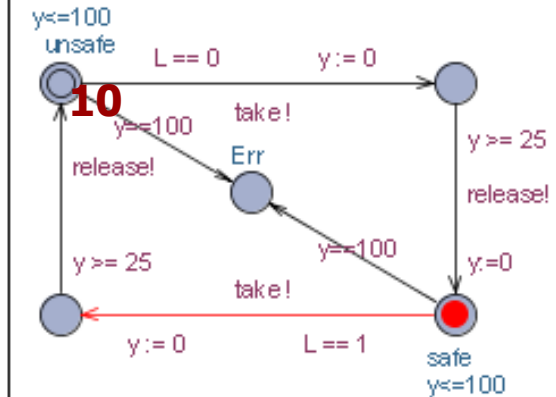
Citroen



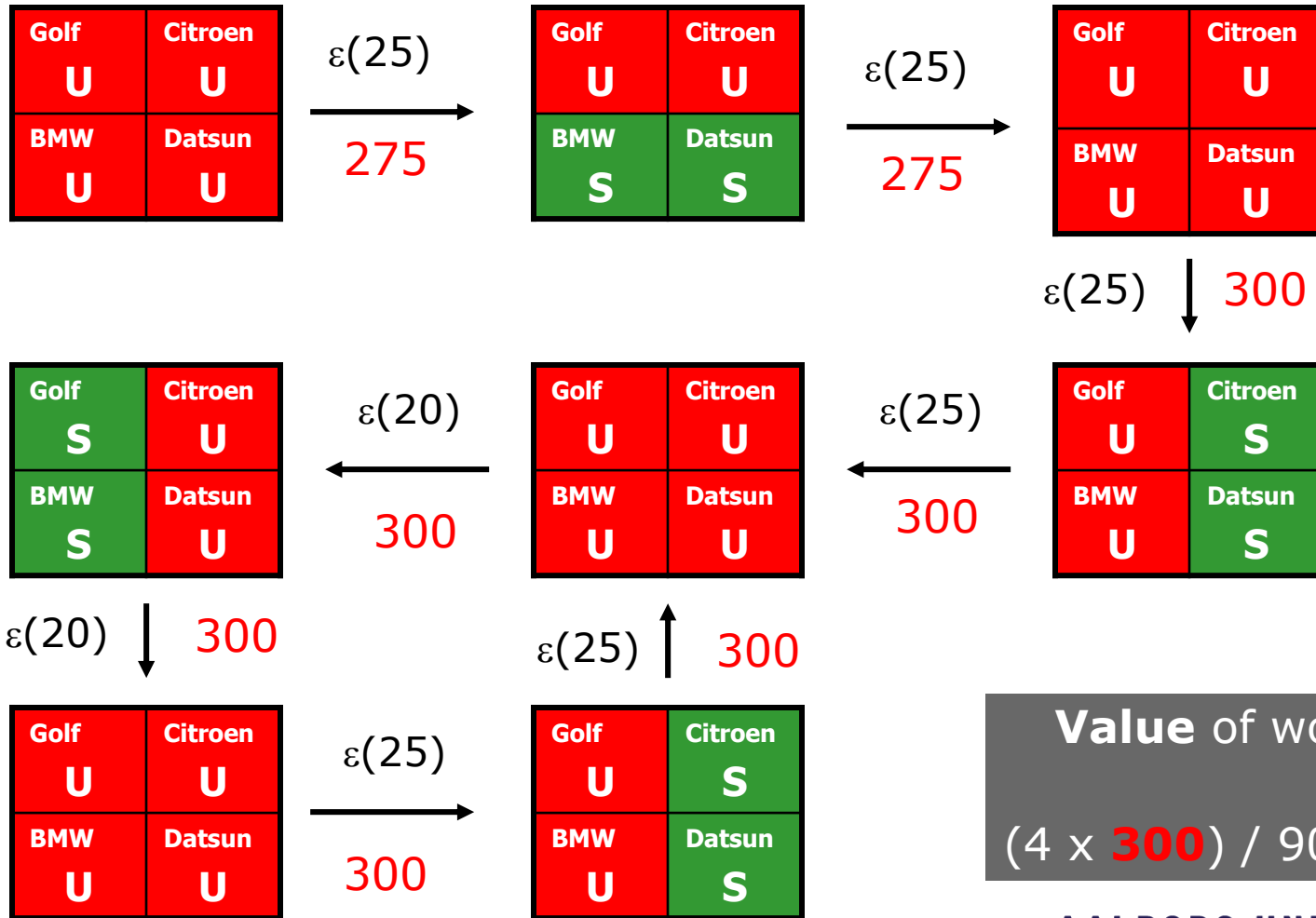
BMW



Datsun



Workplan I



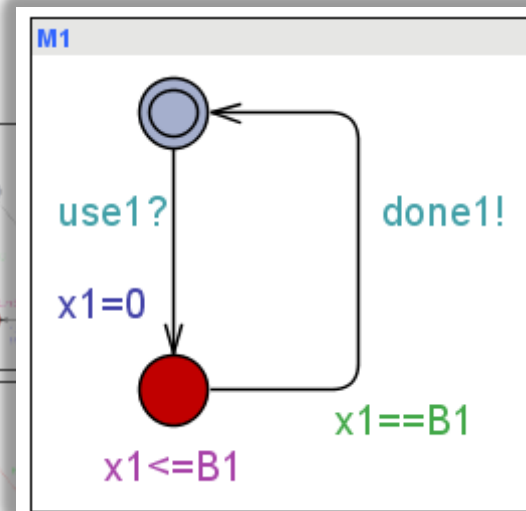
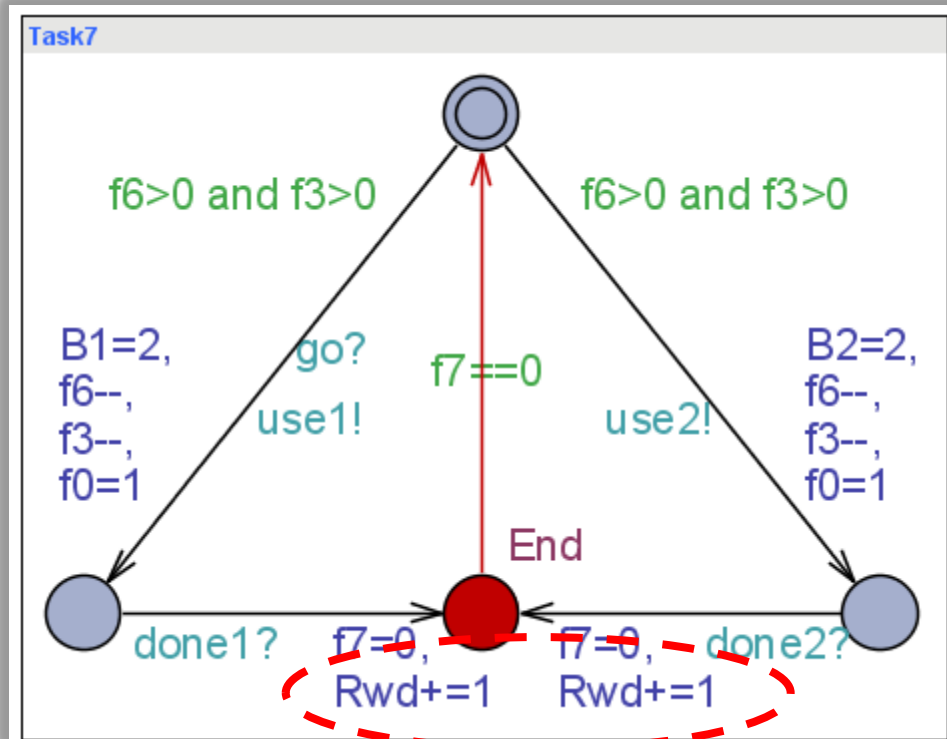
Value of workplan:

$$(4 \times 300) / 90 = 13.33$$

Workplan II

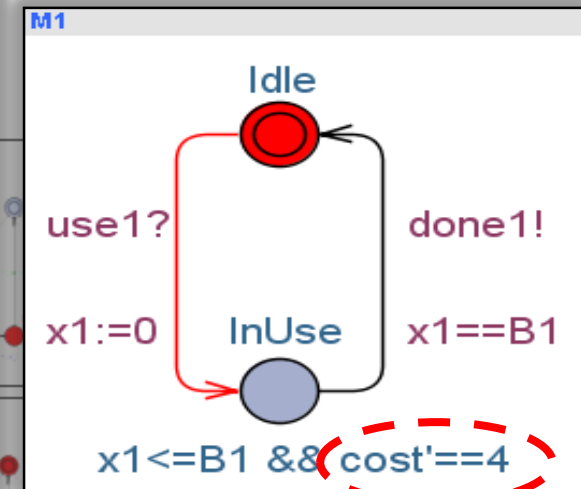
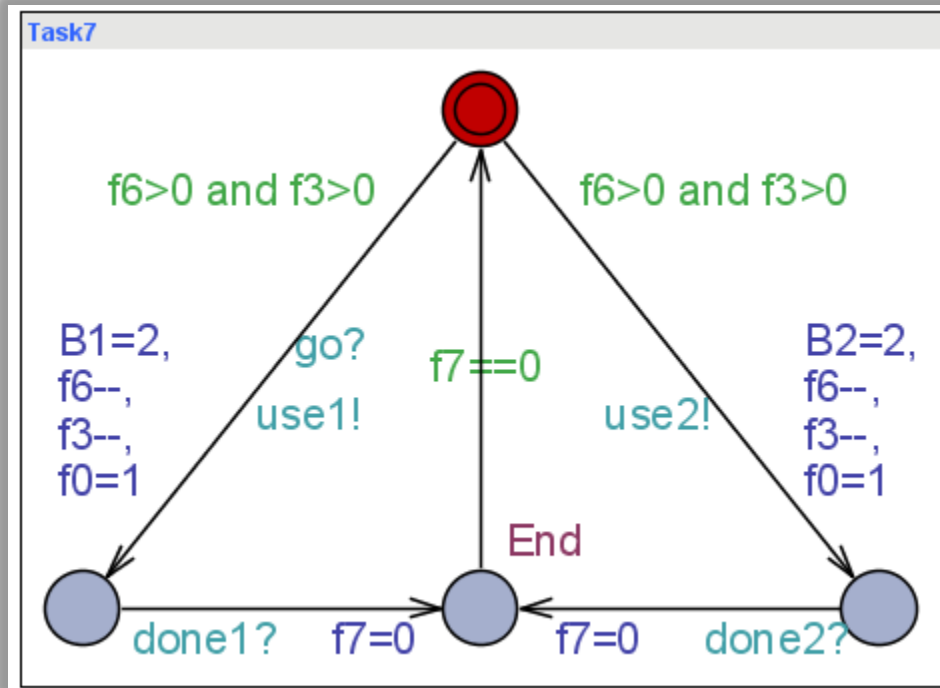


Optimal Infinite Scheduling



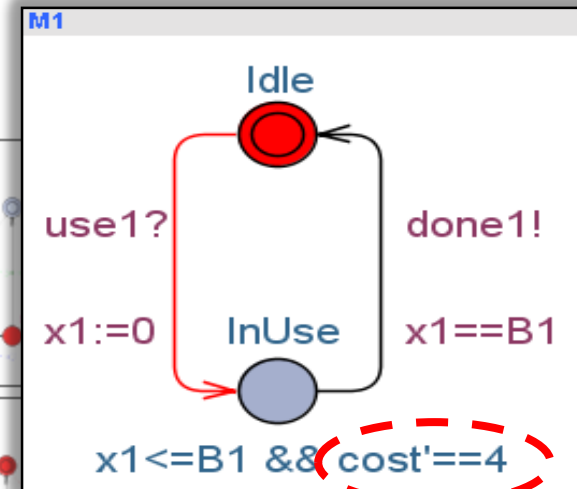
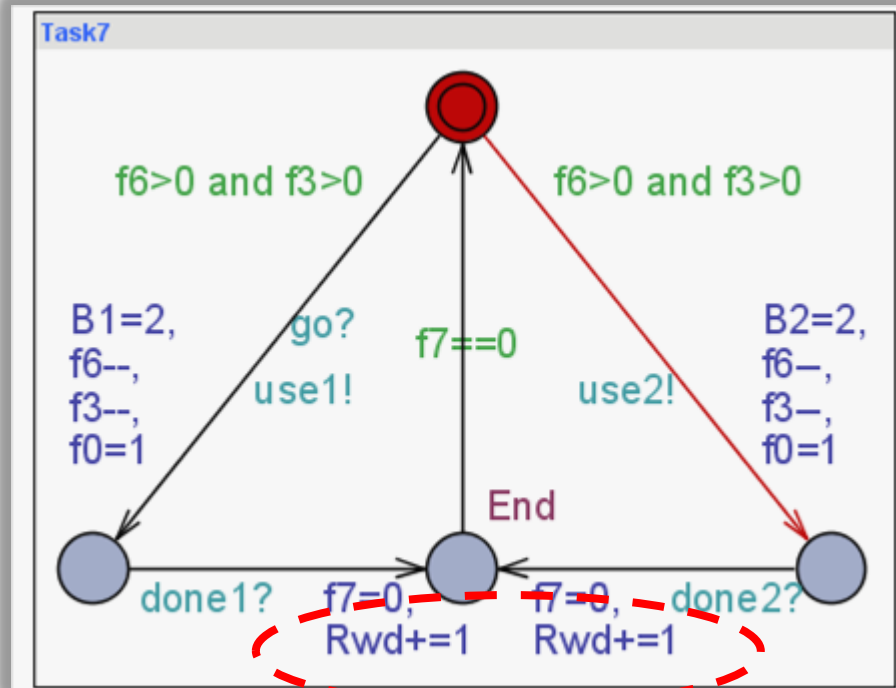
Maximize throughput:
i.e. maximize **Reward** / Time in the long run!

Optimal Infinite Scheduling



Minimize Energy Consumption:
i.e. minimize **Cost** / Time in the long run

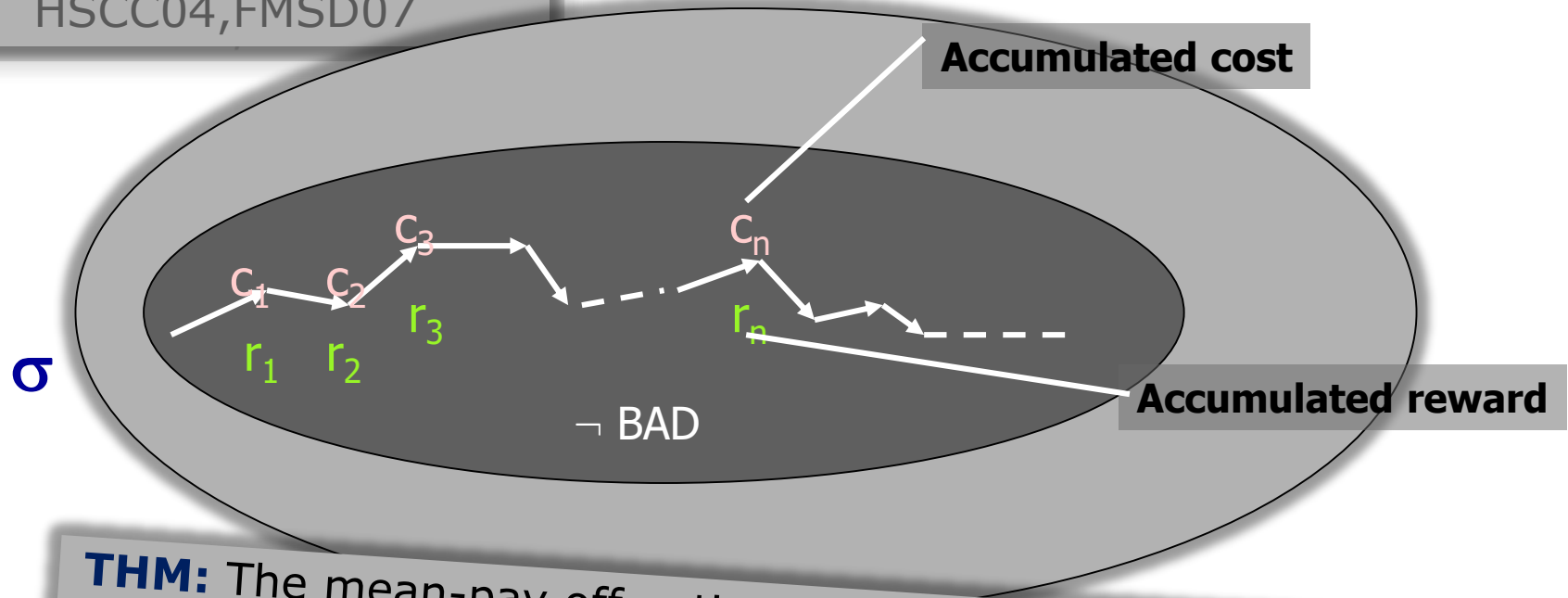
Optimal Infinite Scheduling



Maximize throughput:
i.e. maximize **Reward** / **Cost** in the long run

Mean Pay-Off Optimality

Bouyer, Brinksma, Larsen:
HSCC04, FMSD07



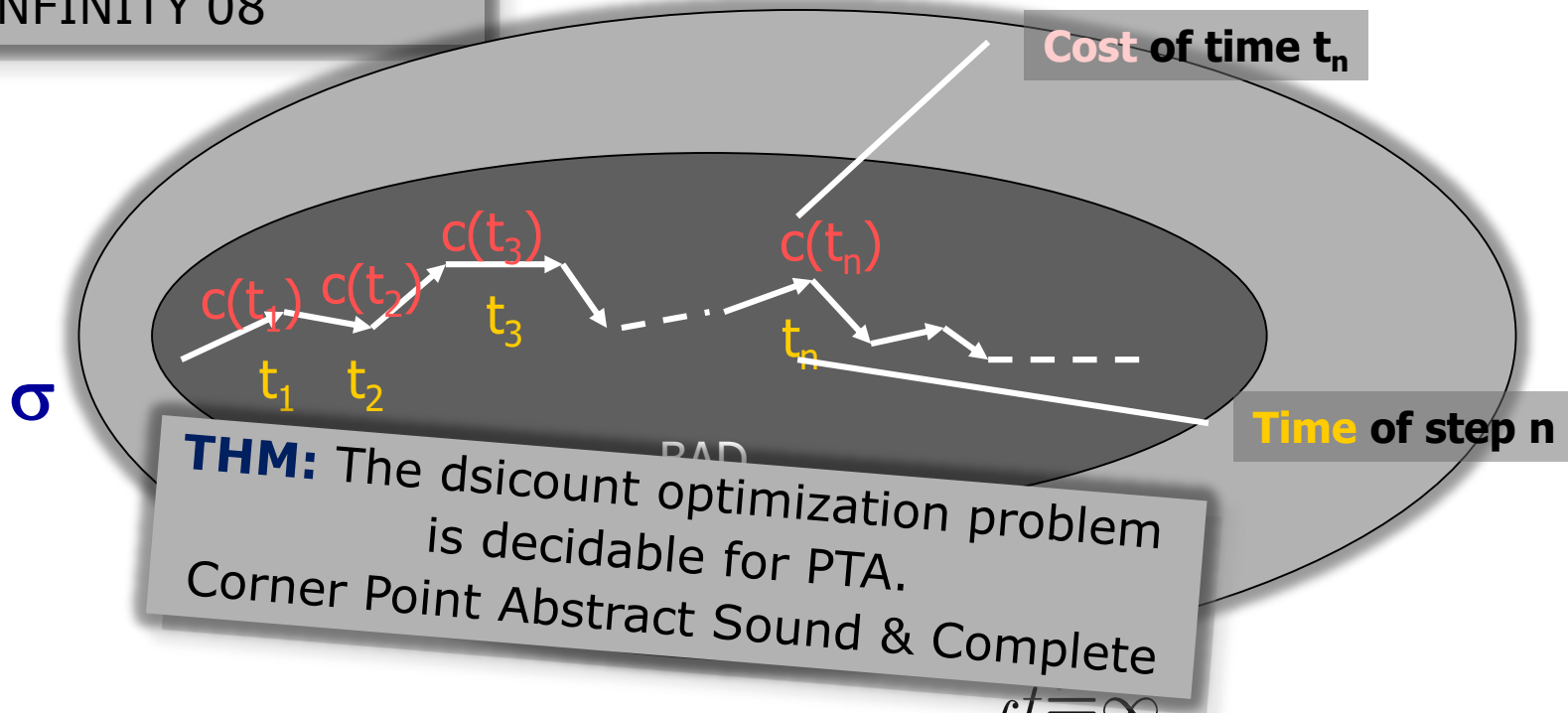
THM: The mean-pay off optimization problem is decidable
(and PSPACE-complete) for PTA.
Corner Point Abstract Sound & Complete

Optimal Schedule σ^* : $\text{val}(\sigma^*) = \inf_{\sigma} \text{val}(\sigma)$

Discount Optimality

$\lambda < 1$: discounting factor

Larsen, Fahrenberg:
INFINITY'08



Value of path σ : $\text{val}(\sigma) = \int_{t=0}^{t=\infty} c(t) \lambda^t dt$

Optimal Schedule σ^* : $\text{val}(\sigma^*) = \inf_{\sigma} \text{val}(\sigma)$

Soundness of Corner Point Abstraction

Lemma

Let Z be a (bounded, closed) zone and let f be a (well-defined) function over Z defined by:

$$f : (t_1, \dots, t_n) \mapsto \frac{a_1 t_1 + \dots + a_n t_n + a}{c_1 t_1 + \dots + c_n t_n + d}$$

then $\inf_Z f$ is obtained at a corner-point of Z (with integer coefficients).

Lemma

Let Z be a (bounded, closed) zone and let f be a function over Z defined by:

$$f : (t_1, \dots, t_n) \mapsto a_1 \lambda^{t_1} + \dots + a_n \lambda^{t_n} + a$$

then $\inf_Z f$ is obtained at a corner-point of Z (with integer coefficients).



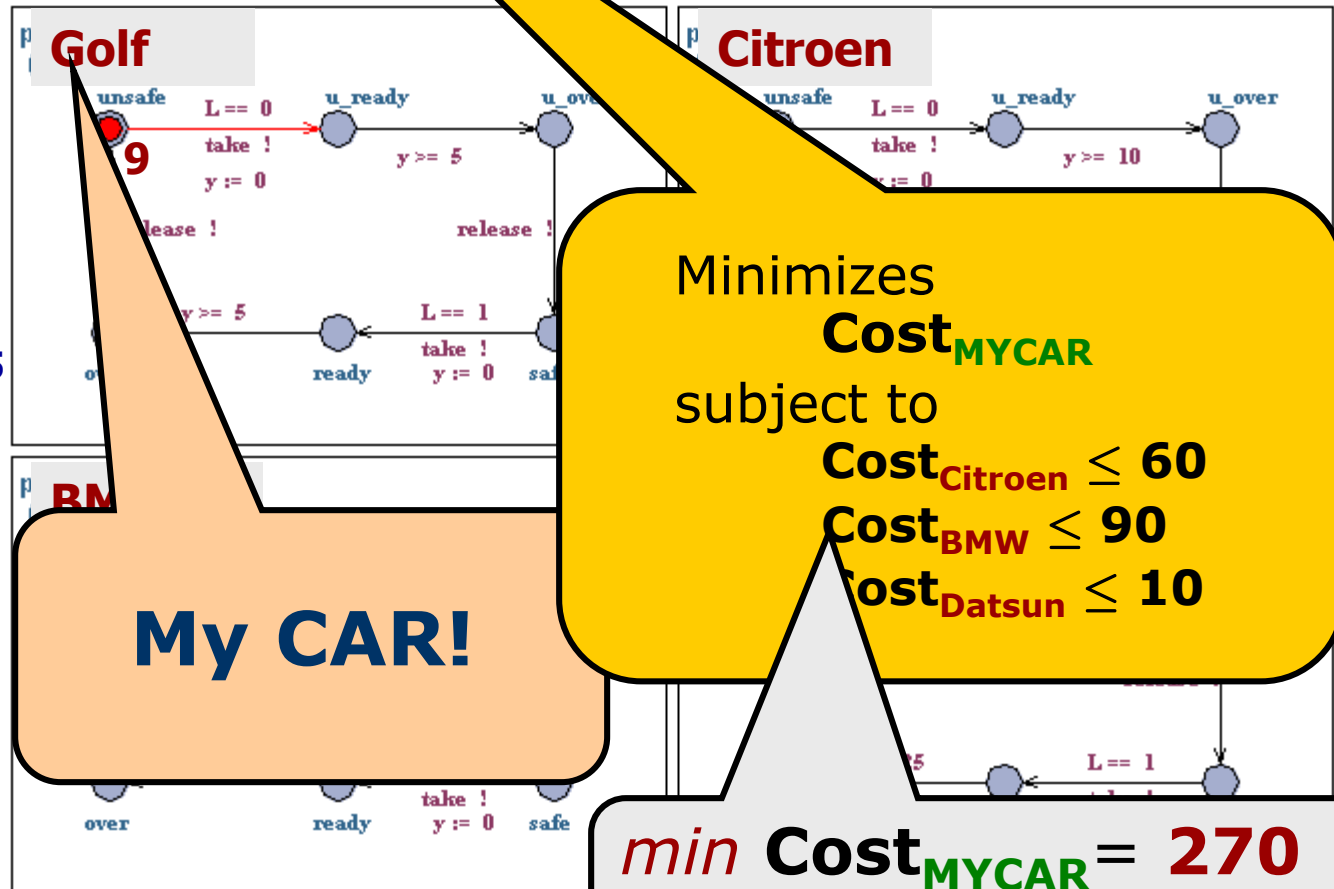
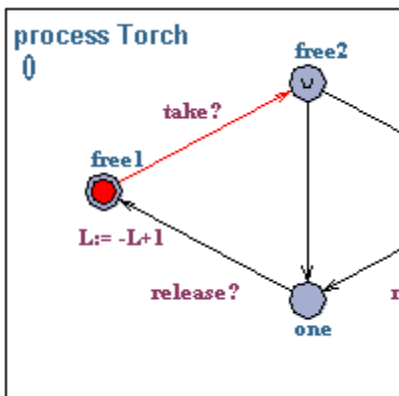
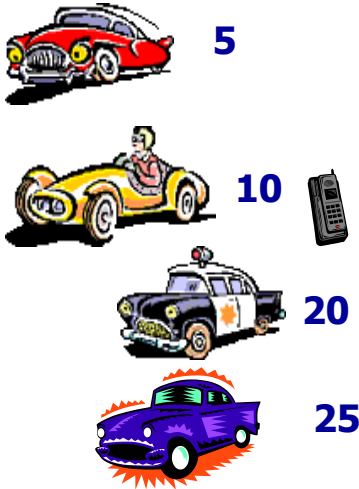
Optimal Conditional Reachability

with Jacob I. Rasmussen

CONDITIONAL

EXAMPLE: Optimal rescue plan for cars with different subscription rates for city driving !

UNSAFE



Minimizes
Cost_{MYCAR}
subject to

$$\text{Cost}_{\text{Citroen}} \leq 60$$

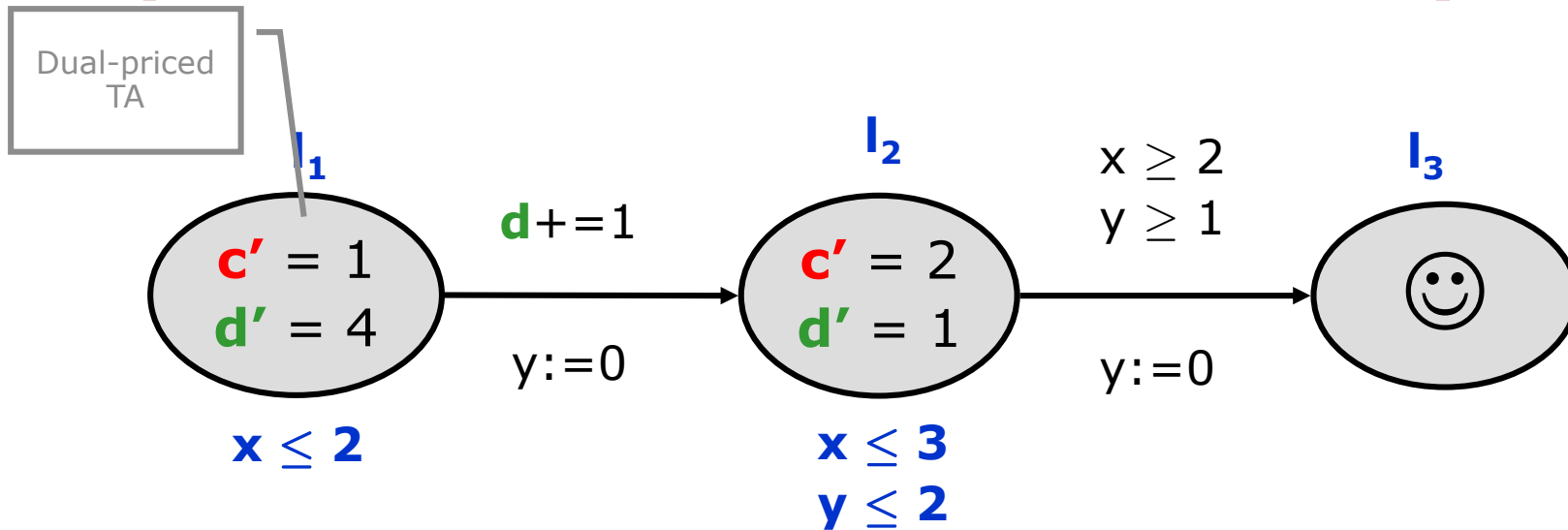
$$\text{Cost}_{\text{BMW}} \leq 90$$

$$\text{Cost}_{\text{Datsun}} \leq 10$$

$$\min \text{Cost}_{\text{MYCAR}} = 270$$

$$\text{time} = 70$$

Optimal *Conditional* Reachability



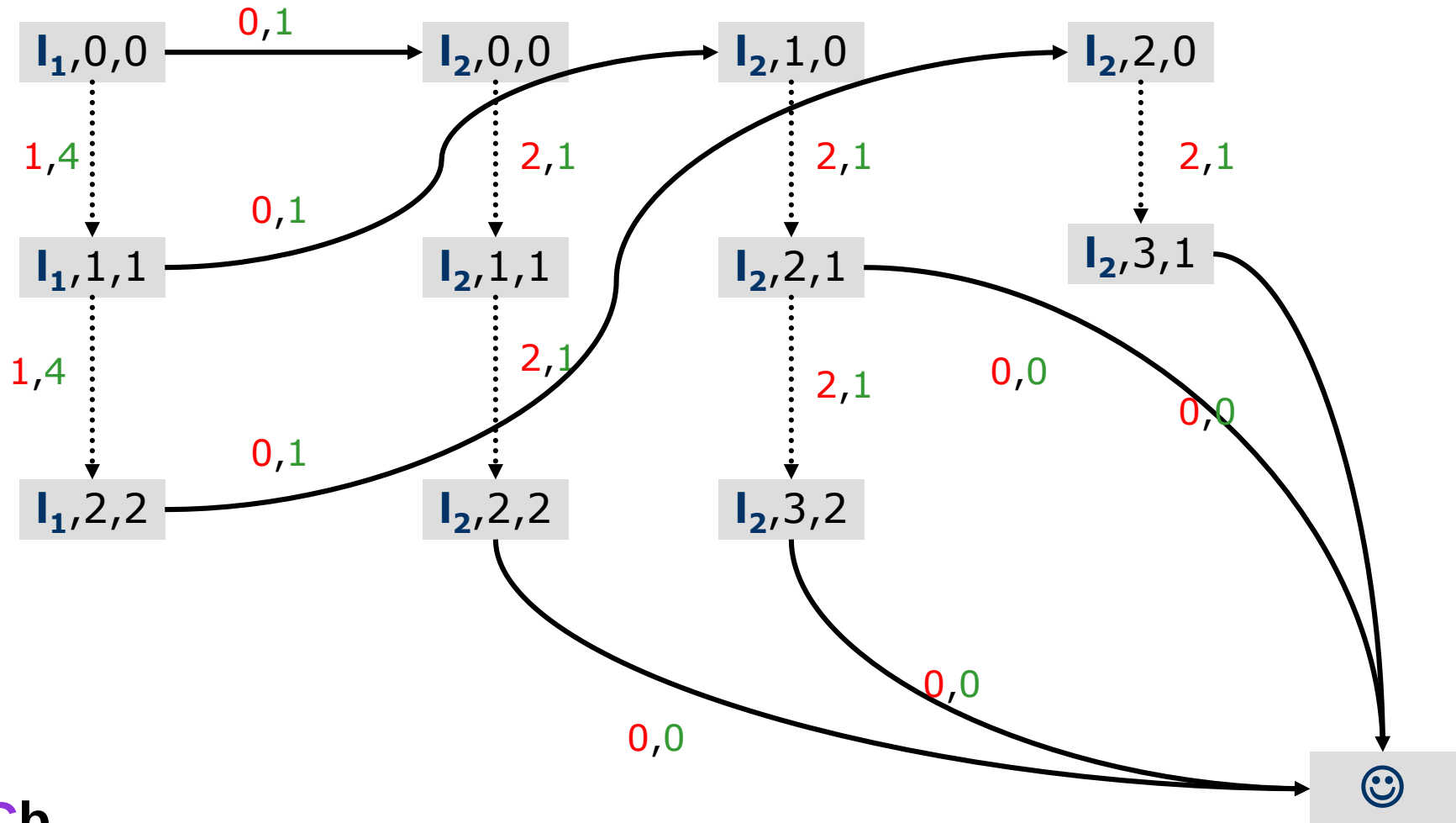
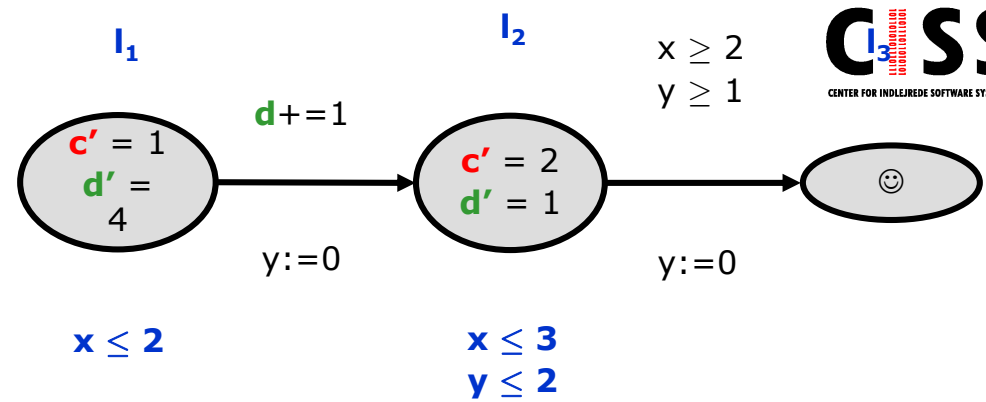
PROBLEM:

Reach l_3 in a way which minimizes c subject to $d \leq 4$

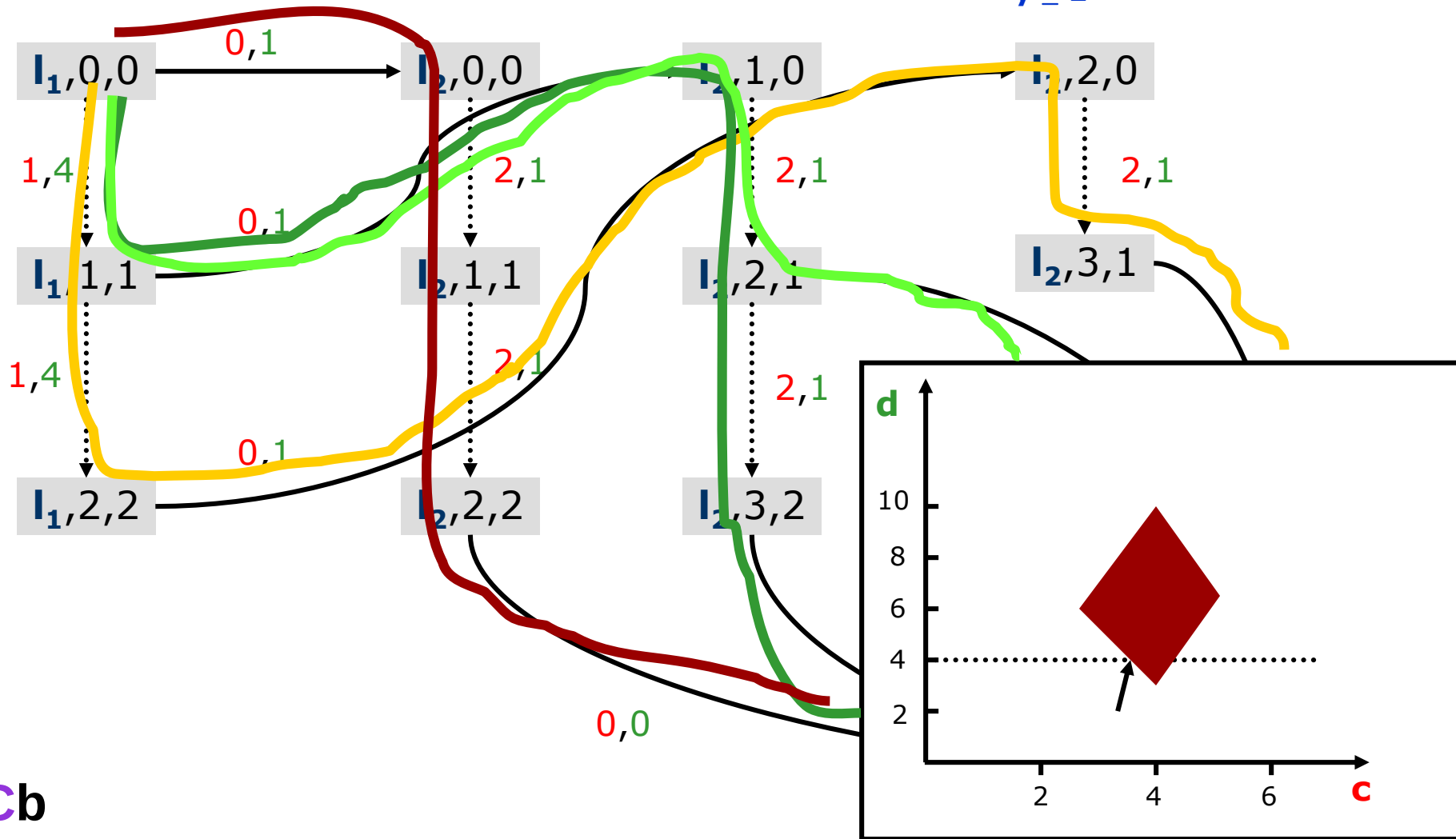
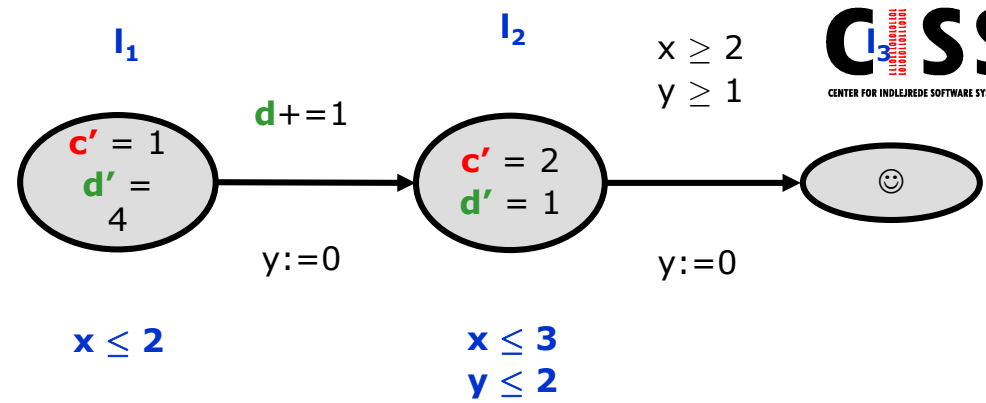
SOLUTION:

$c = 11/3 \rightarrow$
wait $1/3$ in l_1 ; goto l_2 ;
wait $5/3$ in l_2 ; goto l_3

Discrete Trajectories



Discrete Trajectories



Multiple Objective Scheduling

