

Approximation of the value in a weighted timed game

Patricia Bouyer-Decitre

LSV, CNRS & ENS Cachan, France

Joint work with Samy Jaziri and Nicolas Markey



An example: The task graph scheduling problem

Compute $D \times (C \times (A + B)) + (A + B) + (C \times D)$ using two processors:

P_1 (fast):



time	
+	2 picoseconds
×	3 picoseconds

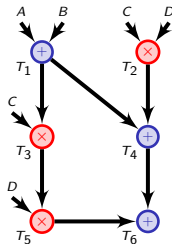
energy	
idle	10 Watt
in use	90 Watts

P_2 (slow):



time	
+	5 picoseconds
×	7 picoseconds

energy	
idle	20 Watts
in use	30 Watts



An example: The task graph scheduling problem

Compute $D \times (C \times (A+B)) + (A+B) + (C \times D)$ using two processors:

P_1 (fast):



time	
+	2 picoseconds
×	3 picoseconds

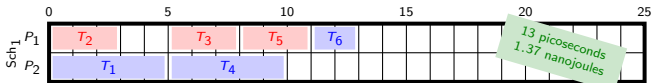
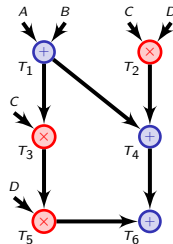
energy	
idle	10 Watt
in use	90 Watts

P_2 (slow):



time	
+	5 picoseconds
×	7 picoseconds

energy	
idle	20 Watts
in use	30 Watts



An example: The task graph scheduling problem

Compute $D \times (C \times (A+B)) + (A+B) + (C \times D)$ using two processors:

P_1 (fast):



time	
+	2 picoseconds
×	3 picoseconds

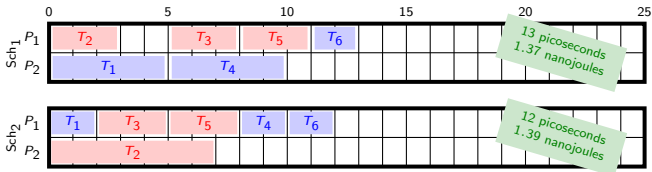
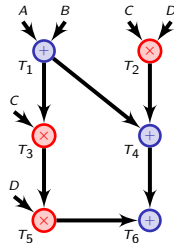
energy	
idle	10 Watt
in use	90 Watts

P_2 (slow):



time	
+	5 picoseconds
×	7 picoseconds

energy	
idle	20 Watts
in use	30 Watts



An example: The task graph scheduling problem

Compute $D \times (C \times (A+B)) + (A+B) + (C \times D)$ using two processors:

P_1 (fast):



time	
+	2 picoseconds
×	3 picoseconds

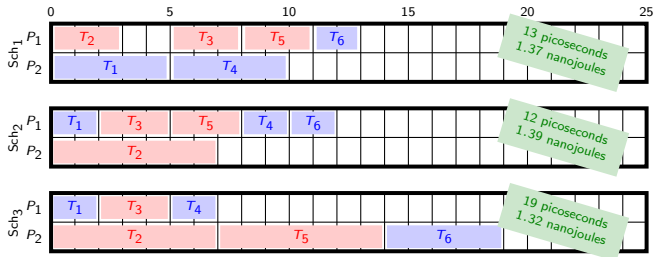
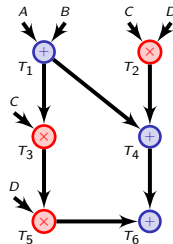
energy	
idle	10 Watt
in use	90 Watts

P_2 (slow):

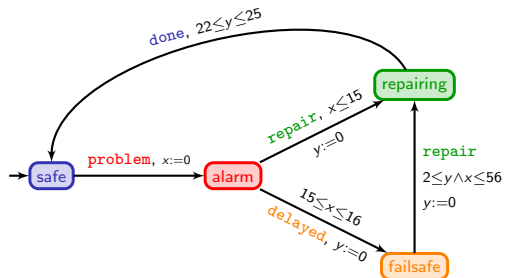


time	
+	5 picoseconds
×	7 picoseconds

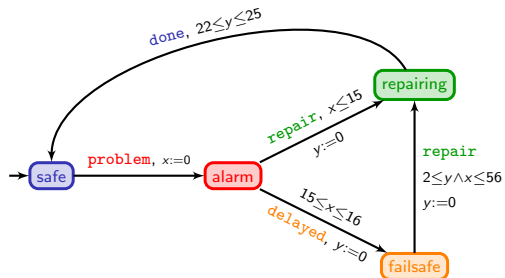
energy	
idle	20 Watts
in use	30 Watts



The model of timed automata



The model of timed automata

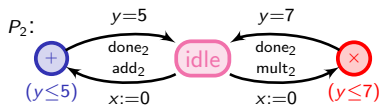
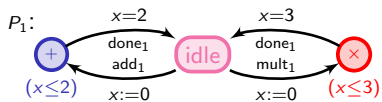


	safe	$\xrightarrow{23}$	safe	$\xrightarrow{\text{problem}}$	alarm	$\xrightarrow{15.6}$	alarm	$\xrightarrow{\text{delayed}}$	failsafe	...
x	0		23		0		15.6		15.6	...
y	0		23		23		38.6		0	
	failsafe	$\xrightarrow{2.3}$	failsafe	$\xrightarrow{\text{repair}}$	repairing	$\xrightarrow{22.1}$	repairing	$\xrightarrow{\text{done}}$	safe	
...	15.6		17.9		17.9		40		40	
	0		2.3		0		22.1		22.1	

Modelling the task graph scheduling problem

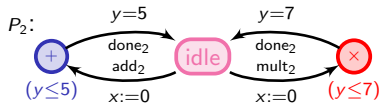
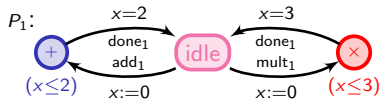
Modelling the task graph scheduling problem

- Processors

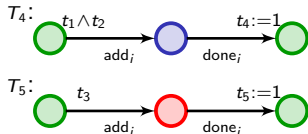


Modelling the task graph scheduling problem

- Processors



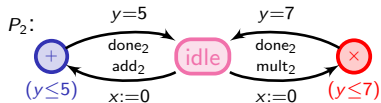
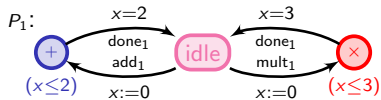
- Tasks



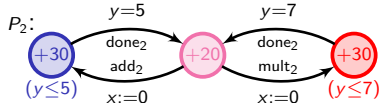
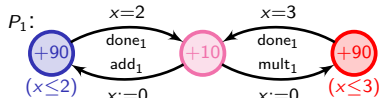
A schedule is a path in the product automaton

Modelling the task graph scheduling problem

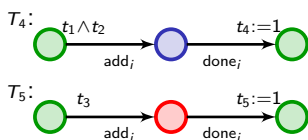
- Processors



- Modelling energy



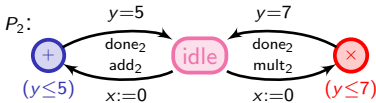
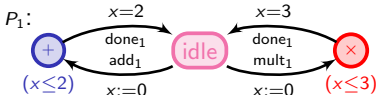
- Tasks



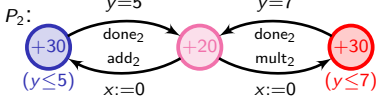
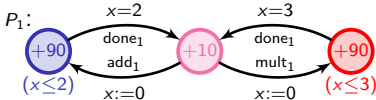
A good schedule is a path in the product automaton with a low cost

Modelling the task graph scheduling problem

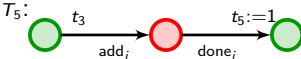
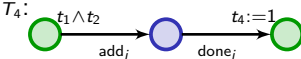
- Processors



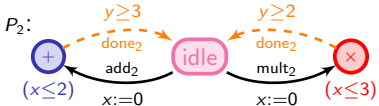
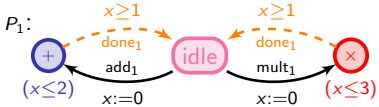
- Modelling energy



- Tasks

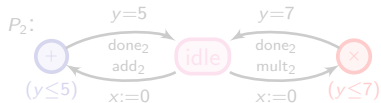
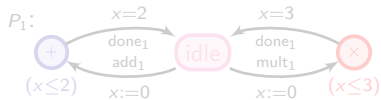


- Modelling uncertainty

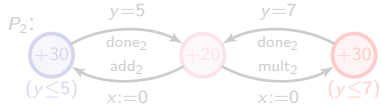
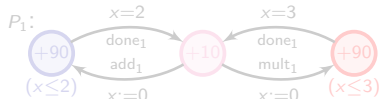


Modelling the task graph scheduling problem

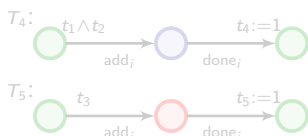
- Processors



- Modelling energy

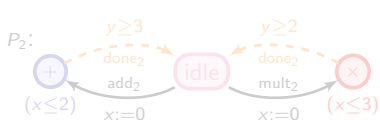
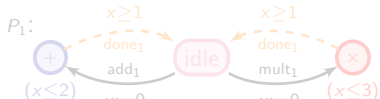


- Tasks

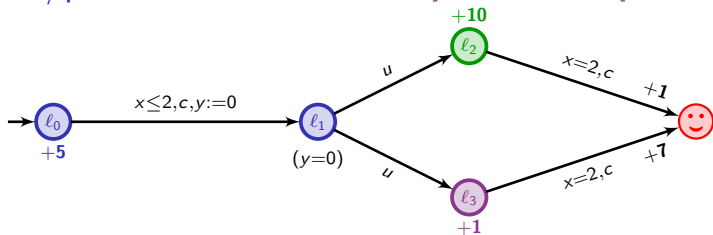


A (good) schedule is a strategy in the product game (with a low cost)

- Modelling uncertainty



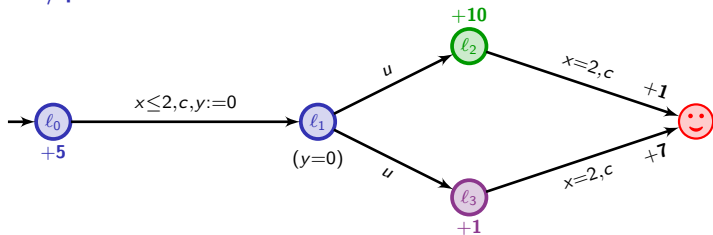
Weighted/priced timed automata [ALP01,BFH+01]



[ALP01] Alur, La Torre, Pappas. Optimal paths in weighted timed automata (*HSCC'01*).

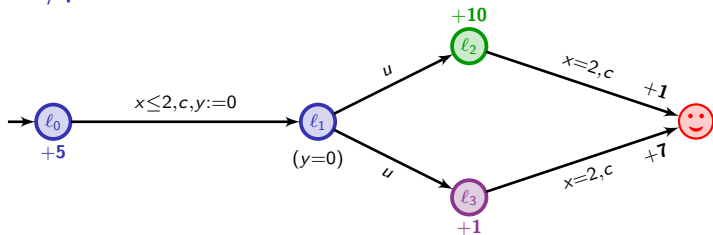
[BFH+01] Behrmann, Fehnker, Hune, Larsen, Pettersson, Romijn, Vaandrager. Minimum-cost reachability in priced timed automata (*HSCC'01*).

Weighted/priced timed automata



	l_0	$\xrightarrow{1.3}$	l_0	\xrightarrow{c}	l_1	\xrightarrow{u}	l_3	$\xrightarrow{0.7}$	l_3	\xrightarrow{c}	😊
x	0		1.3		1.3		1.3		2		
y	0		1.3		0		0		0.7		

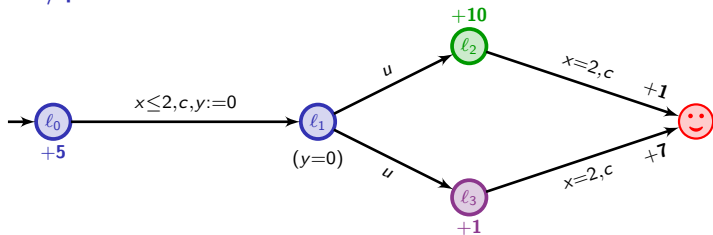
Weighted/priced timed automata



	l_0	$\xrightarrow{1.3}$	l_0	\xrightarrow{c}	l_1	\xrightarrow{u}	l_3	$\xrightarrow{0.7}$	l_3	\xrightarrow{c}	😊
x	0		1.3		1.3		1.3		2		
y	0		1.3		0		0		0.7		

cost :

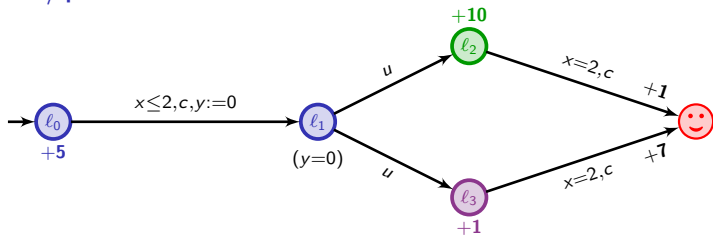
Weighted/priced timed automata



	l_0	$\xrightarrow{1.3}$	l_0	\xrightarrow{c}	l_1	\xrightarrow{u}	l_3	$\xrightarrow{0.7}$	l_3	\xrightarrow{c}	😊
x	0		1.3		1.3		1.3		2		
y	0		1.3		0		0		0.7		

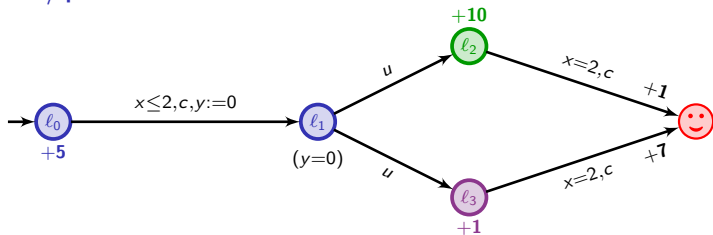
cost : 6.5

Weighted/priced timed automata



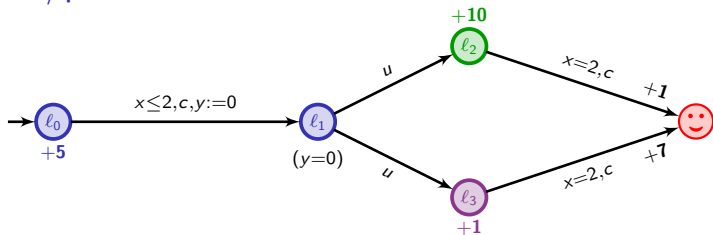
	l_0	$\xrightarrow{1.3}$	l_0	\xrightarrow{c}	l_1	\xrightarrow{u}	l_3	$\xrightarrow{0.7}$	l_3	\xrightarrow{c}	
x :	0		1.3		1.3		1.3		2		
y :	0		1.3		0		0		0.7		
cost :	6.5	+	0								

Weighted/priced timed automata



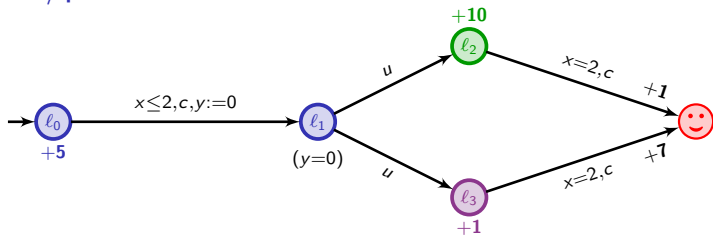
	l_0	$\xrightarrow{1.3}$	l_0	\xrightarrow{c}	l_1	\xrightarrow{u}	l_3	$\xrightarrow{0.7}$	l_3	\xrightarrow{c}	😊
x :	0		1.3		1.3		1.3		2		
y :	0		1.3		0		0		0.7		
cost :	6.5	+	0	+	0						

Weighted/priced timed automata



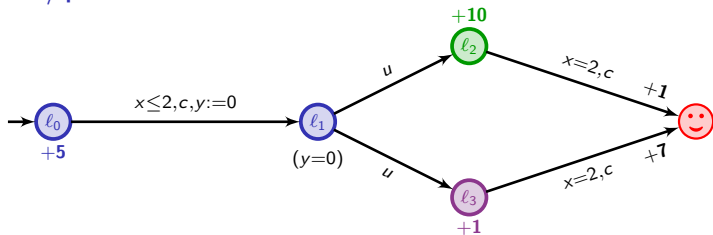
	l_0	$\xrightarrow{1.3}$	l_0	\xrightarrow{c}	l_1	\xrightarrow{u}	l_3	$\xrightarrow{0.7}$	l_3	\xrightarrow{c}	😊
x :	0		1.3		1.3		1.3		2		
y :	0		1.3		0		0		0.7		
cost :		6.5	+	0	+	0	+	0.7			

Weighted/priced timed automata



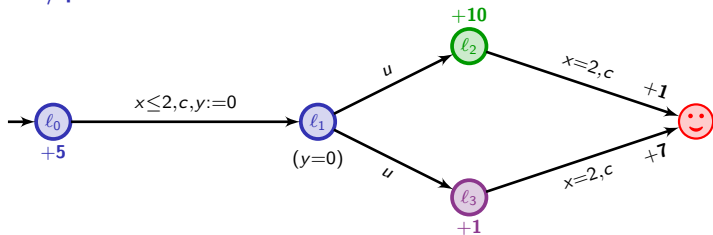
	l_0	$\xrightarrow{1.3}$	l_0	\xrightarrow{c}	l_1	\xrightarrow{u}	l_3	$\xrightarrow{0.7}$	l_3	\xrightarrow{c}	😊
x :	0		1.3		1.3		1.3		2		
y :	0		1.3		0		0		0.7		
cost :		6.5	+	0	+	0	+	0.7	+	7	

Weighted/priced timed automata



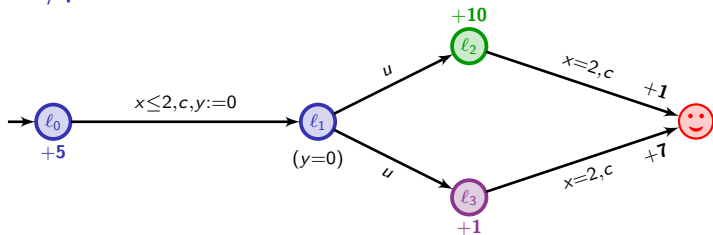
	l_0	$\xrightarrow{1.3}$	l_0	\xrightarrow{c}	l_1	\xrightarrow{u}	l_3	$\xrightarrow{0.7}$	l_3	\xrightarrow{c}	
x :	0		1.3		1.3		1.3		2		
y :	0		1.3		0		0		0.7		
cost :	6.5	+	0	+	0	+	0.7	+	7	=	14.2

Weighted/priced timed automata



Question: what is the optimal cost for reaching 😊?

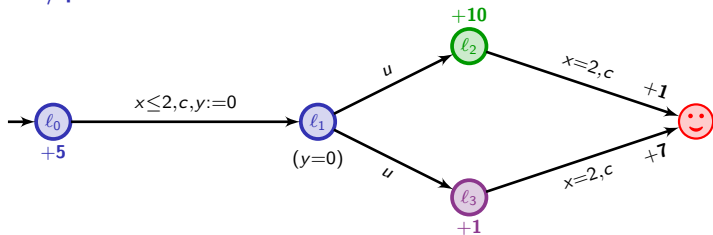
Weighted/priced timed automata



Question: what is the optimal cost for reaching 😊?

$$5t + 10(2 - t) + 1$$

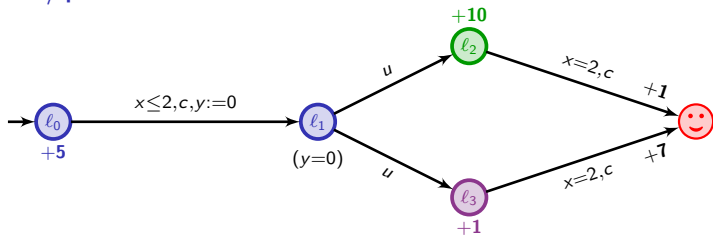
Weighted/priced timed automata



Question: what is the optimal cost for reaching 😊?

$$5t + 10(2 - t) + 1, \quad 5t + (2 - t) + 7$$

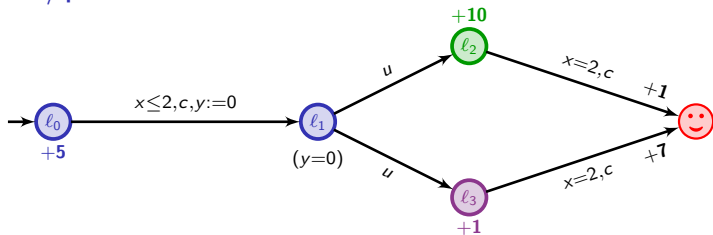
Weighted/priced timed automata



Question: what is the optimal cost for reaching 😊?

$$\min (5t + 10(2 - t) + 1 , 5t + (2 - t) + 7)$$

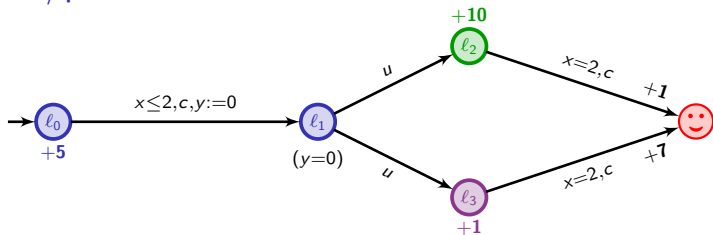
Weighted/priced timed automata



Question: what is the optimal cost for reaching 😊?

$$\inf_{0 \leq t \leq 2} \min (5t + 10(2 - t) + 1 , 5t + (2 - t) + 7) = 9$$

Weighted/priced timed automata

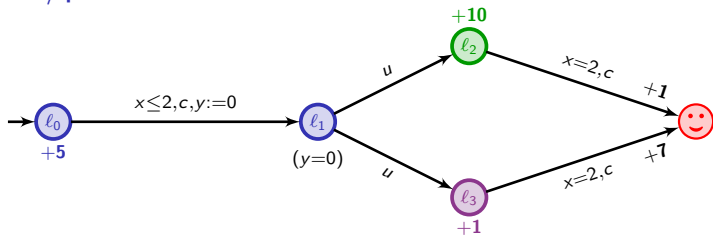


Question: what is the optimal cost for reaching 😊?

$$\inf_{0 \leq t \leq 2} \min (5t + 10(2 - t) + 1 , 5t + (2 - t) + 7) = 9$$

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

Weighted/priced timed automata



Question: what is the optimal cost for reaching 😊?

$$\inf_{0 \leq t \leq 2} \min (5t + 10(2 - t) + 1 , 5t + (2 - t) + 7) = 9$$

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

That can be generalized!

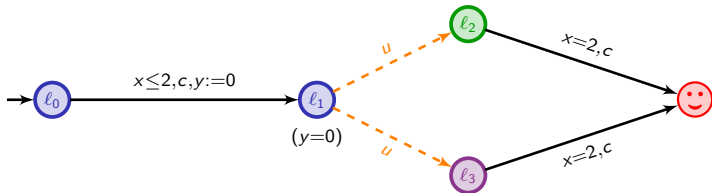
[ALP01] Alur, La Torre, Pappas. Optimal paths in weighted timed automata (*HSCC'01*).

[BFH+01] Behrmann, Fehnker, Hune, Larsen, Pettersson, Romijn, Vaandrager. Minimum-cost reachability in priced timed automata (*HSCC'01*).

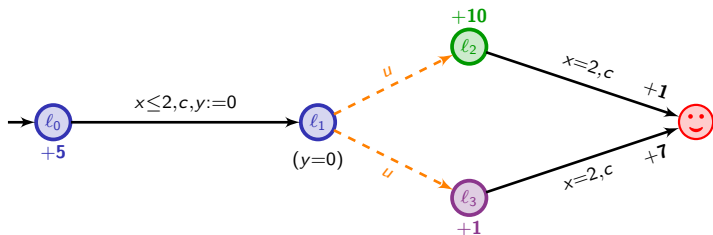
[BBBR07] Bouyer, Brihaye, Bruyère, Raskin. On the optimal reachability problem (*Formal Methods in System Design*).

A simple

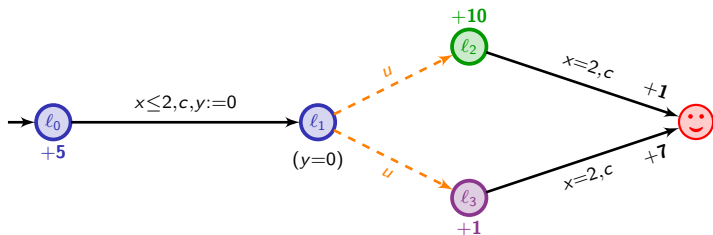
timed game



A simple weighted timed game

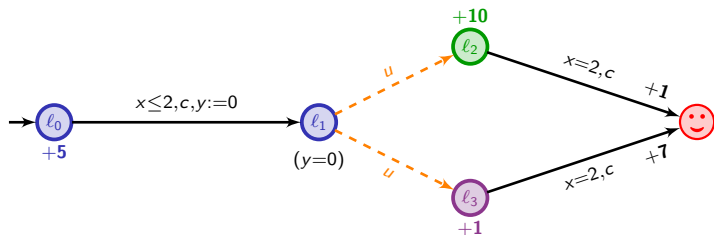


A simple weighted timed game



Question: what is the optimal cost we can ensure while reaching 😊?

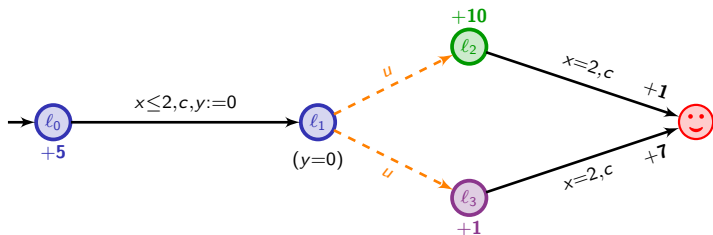
A simple weighted timed game



Question: what is the optimal cost we can ensure while reaching 😊?

$$5t + 10(2 - t) + 1$$

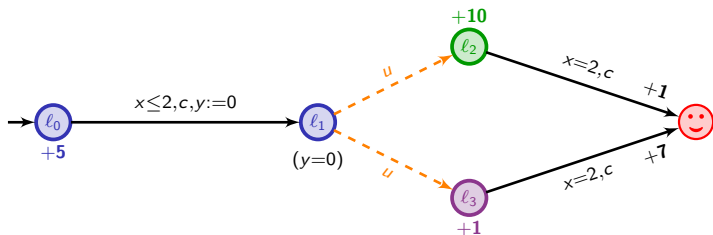
A simple weighted timed game



Question: what is the optimal cost we can ensure while reaching 😊?

$$5t + 10(2 - t) + 1, \quad 5t + (2 - t) + 7$$

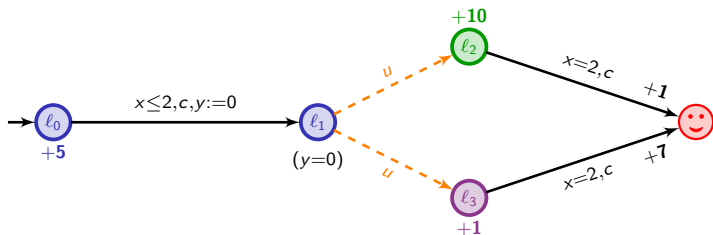
A simple weighted timed game



Question: what is the optimal cost we can ensure while reaching 😊?

$$\max (5t + 10(2 - t) + 1 , 5t + (2 - t) + 7)$$

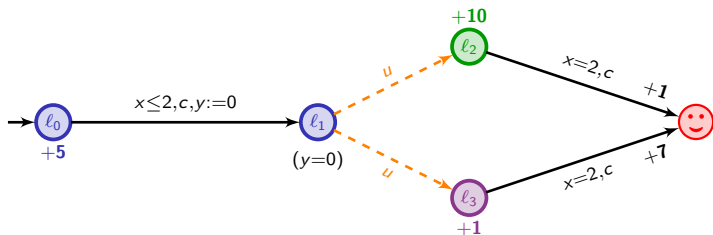
A simple weighted timed game



Question: what is the optimal cost we can ensure while reaching 😊?

$$\inf_{0 \leq t \leq 2} \max (5t + 10(2 - t) + 1 , 5t + (2 - t) + 7) = 14 + \frac{1}{3}$$

A simple weighted timed game



Question: what is the optimal cost we can ensure while reaching 😊?

$$\inf_{0 \leq t \leq 2} \max (5t + 10(2 - t) + 1 , 5t + (2 - t) + 7) = 14 + \frac{1}{3}$$

~ strategy: wait in l_0 , and when $t = \frac{4}{3}$, go to l_1

Optimal reachability in weighted timed games (1)

This topic has been fairly hot these last fifteen years...

[LMM02,ABM04,BCFL04,BBR05,BBM06,BLMR06,Rut11,HIM13,BGK+14]

[LMM02] La Torre, Mukhopadhyay, Murano. Optimal-reachability and control for acyclic weighted timed automata (*TCS'02*).

[ABM04] Alur, Bernardsky, Madhusudan. Optimal reachability in weighted timed games (*ICALP'04*).

[BCFL04] Bouyer, Cassez, Fleury, Larsen. Optimal strategies in priced timed game automata (*FSTTCS'04*).

[BBR05] Brihaye, Bruyère, Raskin. On optimal timed strategies (*FORMATS'05*).

[BBM06] Bouyer, Brihaye, Markey. Improved undecidability results on weighted timed automata (*Information Processing Letters*).

[BLMR06] Bouyer, Larsen, Markey, Rasmussen. Almost-optimal strategies in one-clock priced timed automata (*FSTTCS'06*).

[Rut11] Rutkowski. Two-player reachability-price games on single-clock timed automata (*QAPL'11*).

[HIM13] Hansen, Ibsen-Jensen, Miltersen. A faster algorithm for solving one-clock priced timed games (*CONCUR'13*).

[BGK+14] Brihaye, Geeraerts, Krishna, Manasa, Monmege, Trivedi. Adding Negative Prices to Priced Timed Games (*CONCUR'14*).

Optimal reachability in weighted timed games (1)

This topic has been fairly hot these last fifteen years...

[LMM02,ABM04,BCFL04,BBR05,BBM06,BLMR06,Rut11,HIM13,BGK+14]

[LMM02]

Tree-like weighted timed games can be solved in 2EXPTIME.

Optimal reachability in weighted timed games (1)

This topic has been fairly hot these last fifteen years...

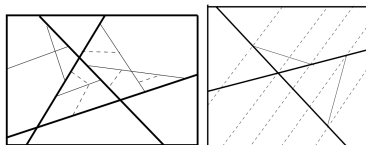
[LMM02,ABM04,BCFL04,BBR05,BBM06,BLMR06,Rut11,HIM13,BGK+14]

[LMM02]

Tree-like weighted timed games can be solved in 2EXPTIME.

[ABM04,BCFL04]

Depth- k weighted timed games can be solved in EXPTIME. There is a symbolic algorithm to solve weighted timed games **with a strongly non-Zeno cost**.



Optimal reachability in weighted timed games (2)

[BBR05, BBM06]

In weighted timed games, the optimal cost **cannot be computed**, as soon as games have three clocks or more.

Optimal reachability in weighted timed games (2)

[BBR05,BBM06]

In weighted timed games, the optimal cost **cannot be computed**, as soon as games have three clocks or more.

[BLMR06,Rut11,HIM13,BGK+14]

Turn-based optimal timed games are **decidable** in EXPTIME (resp. PTIME) when automata have a single clock (resp. with two rates). They are PTIME-hard.

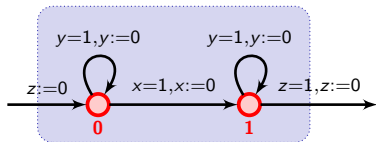
Computing the optimal cost: why is that hard?

Given two clocks x and y , we can check whether $y = 2x$.

Computing the optimal cost: why is that hard?

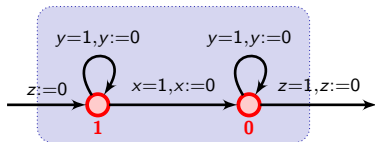
Given two clocks x and y , we can check whether $y = 2x$.

$\text{Add}^+(x)$



The cost is increased by x_0

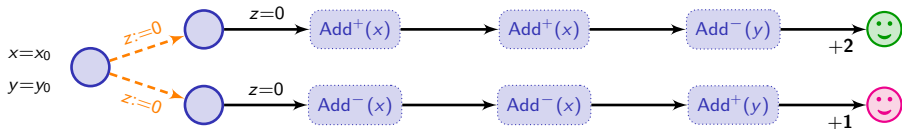
$\text{Add}^-(x)$



The cost is increased by $1-x_0$

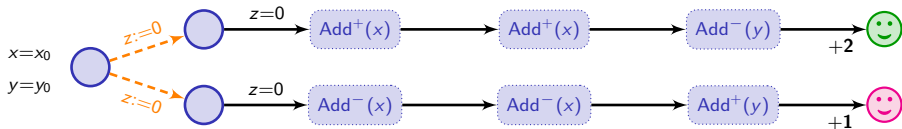
Computing the optimal cost: why is that hard?


Given two clocks x and y , we can check whether $y = 2x$.



Computing the optimal cost: why is that hard?

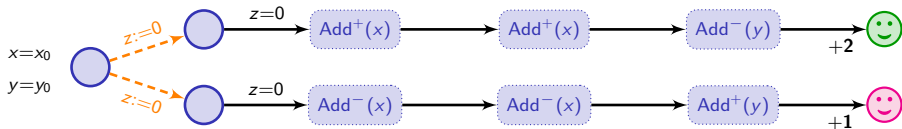
Given two clocks x and y , we can check whether $y = 2x$.





- In , $\text{cost} = 2x_0 + (1 - y_0) + 2$

Computing the optimal cost: why is that hard?

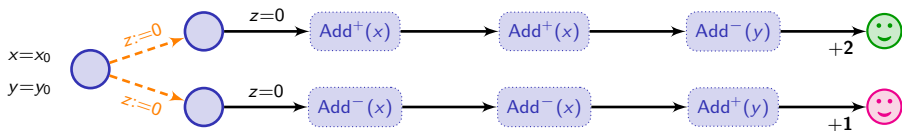
Given two clocks x and y , we can check whether $y = 2x$.





- In , $\text{cost} = 2x_0 + (1 - y_0) + 2$
- In , $\text{cost} = 2(1 - x_0) + y_0 + 1$

Computing the optimal cost: why is that hard?

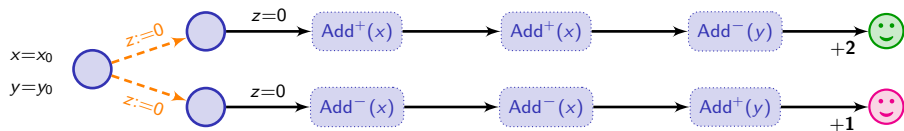
Given two clocks x and y , we can check whether $y = 2x$.





- In , $\text{cost} = 2x_0 + (1 - y_0) + 2$
- In , $\text{cost} = 2(1 - x_0) + y_0 + 1$
- if $y_0 < 2x_0$, **player 2** chooses the first branch: $\text{cost} > 3$

Computing the optimal cost: why is that hard?

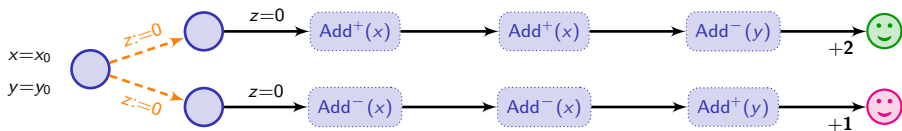
Given two clocks x and y , we can check whether $y = 2x$.





- In , $\text{cost} = 2x_0 + (1 - y_0) + 2$
In , $\text{cost} = 2(1 - x_0) + y_0 + 1$
- if $y_0 < 2x_0$, **player 2** chooses the first branch: $\text{cost} > 3$
if $y_0 > 2x_0$, **player 2** chooses the second branch: $\text{cost} > 3$

Computing the optimal cost: why is that hard?

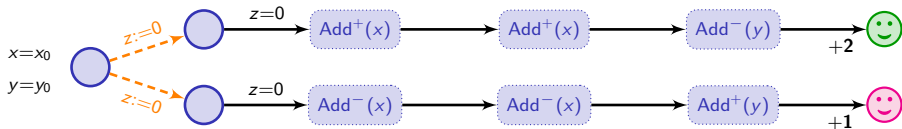
Given two clocks x and y , we can check whether $y = 2x$.





- In , $\text{cost} = 2x_0 + (1 - y_0) + 2$
In , $\text{cost} = 2(1 - x_0) + y_0 + 1$
- if $y_0 < 2x_0$, **player 2** chooses the first branch: $\text{cost} > 3$
if $y_0 > 2x_0$, **player 2** chooses the second branch: $\text{cost} > 3$
if $y_0 = 2x_0$, in both branches, $\text{cost} = 3$

Computing the optimal cost: why is that hard?

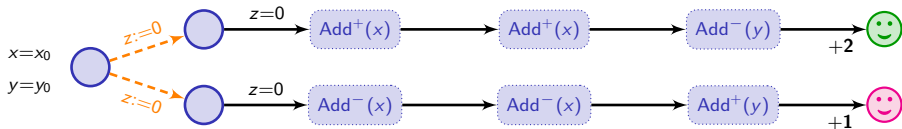
Given two clocks x and y , we can check whether $y = 2x$.





- In , $\text{cost} = 2x_0 + (1 - y_0) + 2$
In , $\text{cost} = 2(1 - x_0) + y_0 + 1$
- if $y_0 < 2x_0$, **player 2** chooses the first branch: $\text{cost} > 3$
if $y_0 > 2x_0$, **player 2** chooses the second branch: $\text{cost} > 3$
if $y_0 = 2x_0$, in both branches, $\text{cost} = 3$
 \leadsto **player 2** can enforce $\text{cost } 3 + |y_0 - 2x_0|$

Computing the optimal cost: why is that hard?

Given two clocks x and y , we can check whether $y = 2x$.



- In , $\text{cost} = 2x_0 + (1 - y_0) + 2$
In , $\text{cost} = 2(1 - x_0) + y_0 + 1$
- if $y_0 < 2x_0$, **player 2** chooses the first branch: $\text{cost} > 3$
if $y_0 > 2x_0$, **player 2** chooses the second branch: $\text{cost} > 3$
if $y_0 = 2x_0$, in both branches, $\text{cost} = 3$
 \leadsto **player 2** can enforce $\text{cost } 3 + |y_0 - 2x_0|$
- Player 1 has a winning strategy with $\text{cost} \leq 3$ iff $y_0 = 2x_0$

Computing the optimal cost: why is that hard?

Player 1 will simulate a two-counter machine:

- each instruction is encoded as a module;
- the counter values c_1 and c_2 are encoded by two clocks:

$$x = \frac{1}{2^{c_1}} \quad \text{and} \quad y = \frac{1}{3^{c_2}}$$

Computing the optimal cost: why is that hard?

Player 1 will simulate a two-counter machine:

- each instruction is encoded as a module;
- the counter values c_1 and c_2 are encoded by two clocks:

$$x = \frac{1}{2^{c_1}} \quad \text{and} \quad y = \frac{1}{3^{c_2}}$$

The two-counter machine has a halting computation iff player 1 has a winning strategy to ensure a cost no more than 3.

Computing the optimal cost: why is that hard?

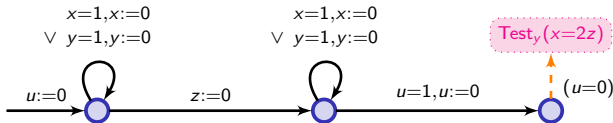
Player 1 will simulate a two-counter machine:

- each instruction is encoded as a module;
- the counter values c_1 and c_2 are encoded by two clocks:

$$x = \frac{1}{2^{c_1}} \quad \text{and} \quad y = \frac{1}{3^{c_2}}$$

The two-counter machine has a halting computation iff player 1 has a winning strategy to ensure a cost no more than 3.

Globally, $(x \leq 1, y \leq 1, u \leq 1)$



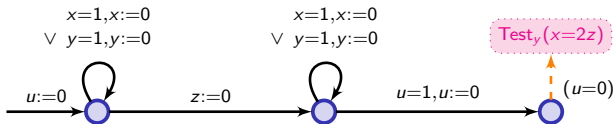
Computing the optimal cost: why is that hard?

Player 1 will simulate a two-counter machine:

- each instruction is encoded as a module;
- the counter values c_1 and c_2 are encoded by two clocks:

$$x = \frac{1}{2^{c_1}} \quad \text{and} \quad y = \frac{1}{3^{c_2}}$$

The two-counter machine has a halting computation iff player 1 has a winning strategy to ensure a cost no more than 3.



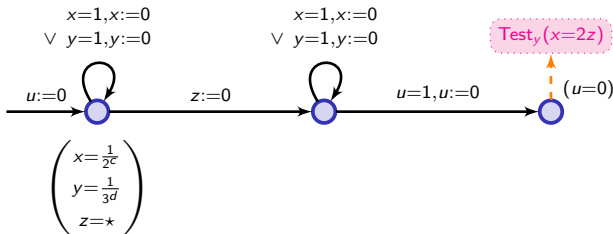
Computing the optimal cost: why is that hard?

Player 1 will simulate a two-counter machine:

- each instruction is encoded as a module;
- the counter values c_1 and c_2 are encoded by two clocks:

$$x = \frac{1}{2^{c_1}} \quad \text{and} \quad y = \frac{1}{3^{c_2}}$$

The two-counter machine has a halting computation iff player 1 has a winning strategy to ensure a cost no more than 3.



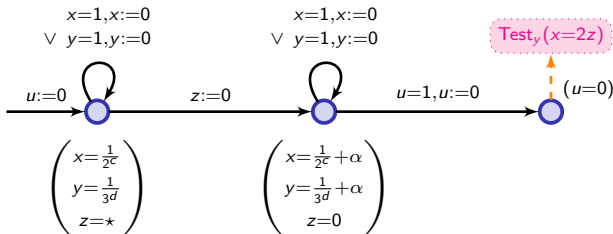
Computing the optimal cost: why is that hard?

Player 1 will simulate a two-counter machine:

- each instruction is encoded as a module;
- the counter values c_1 and c_2 are encoded by two clocks:

$$x = \frac{1}{2^{c_1}} \quad \text{and} \quad y = \frac{1}{3^{c_2}}$$

The two-counter machine has a halting computation iff player 1 has a winning strategy to ensure a cost no more than 3.



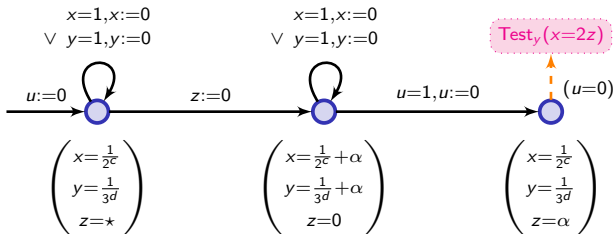
Computing the optimal cost: why is that hard?

Player 1 will simulate a two-counter machine:

- each instruction is encoded as a module;
- the counter values c_1 and c_2 are encoded by two clocks:

$$x = \frac{1}{2c_1} \quad \text{and} \quad y = \frac{1}{3c_2}$$

The two-counter machine has a halting computation iff player 1 has a winning strategy to ensure a cost no more than 3.



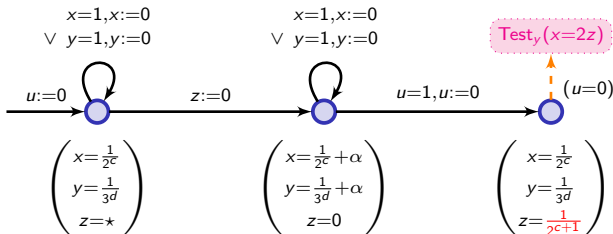
Computing the optimal cost: why is that hard?

Player 1 will simulate a two-counter machine:

- each instruction is encoded as a module;
- the counter values c_1 and c_2 are encoded by two clocks:

$$x = \frac{1}{2^{c_1}} \quad \text{and} \quad y = \frac{1}{3^{c_2}}$$

The two-counter machine has a halting computation iff player 1 has a winning strategy to ensure a cost no more than 3.



Shape of the reduction

Are we done?

Are we done? No! Let's be a bit more precise!

Are we done? No! Let's be a bit more precise!

Given \mathcal{G} a weighted timed game,

- a strategy σ is winning whenever all its outcomes are winning;

Are we done? No! Let's be a bit more precise!

Given \mathcal{G} a weighted timed game,

- a strategy σ is winning whenever all its outcomes are winning;
- Cost of a winning strategy σ :

$$\text{cost}(\sigma) = \sup\{\text{cost}(\rho) \mid \rho \text{ outcome of } \sigma \text{ up to the target}\}$$

Are we done? No! Let's be a bit more precise!

Given \mathcal{G} a weighted timed game,

- a strategy σ is winning whenever all its outcomes are winning;
- Cost of a winning strategy σ :

$$\text{cost}(\sigma) = \sup\{\text{cost}(\rho) \mid \rho \text{ outcome of } \sigma \text{ up to the target}\}$$

- Optimal cost:

$$\text{optcost}_{\mathcal{G}} = \inf_{\sigma \text{ winning strat.}} \text{cost}(\sigma)$$

(set it to $+\infty$ if there is no winning strategy)

Are we done? No! Let's be a bit more precise!

Given \mathcal{G} a weighted timed game,

- a strategy σ is winning whenever all its outcomes are winning;
- Cost of a winning strategy σ :

$$\text{cost}(\sigma) = \sup\{\text{cost}(\rho) \mid \rho \text{ outcome of } \sigma \text{ up to the target}\}$$

- Optimal cost:

$$\text{optcost}_{\mathcal{G}} = \inf_{\sigma \text{ winning strat.}} \text{cost}(\sigma)$$

(set it to $+\infty$ if there is no winning strategy)

Two problems of interest

- The **value problem** asks, given \mathcal{G} and a threshold $\bowtie c$, whether $\text{optcost}_{\mathcal{G}} \bowtie c$?

Are we done? No! Let's be a bit more precise!

Given \mathcal{G} a weighted timed game,

- a strategy σ is winning whenever all its outcomes are winning;
- Cost of a winning strategy σ :

$$\text{cost}(\sigma) = \sup\{\text{cost}(\rho) \mid \rho \text{ outcome of } \sigma \text{ up to the target}\}$$

- Optimal cost:

$$\text{optcost}_{\mathcal{G}} = \inf_{\sigma \text{ winning strat.}} \text{cost}(\sigma)$$

(set it to $+\infty$ if there is no winning strategy)

Two problems of interest

- The **value problem** asks, given \mathcal{G} and a threshold $\bowtie c$, whether $\text{optcost}_{\mathcal{G}} \bowtie c$?
- The **existence problem** asks, given \mathcal{G} and a threshold $\bowtie c$, whether there exists a winning strategy in \mathcal{G} such that $\text{cost}(\sigma) \bowtie c$?

Are we done? No! Let's be a bit more precise!

Given \mathcal{G} a weighted timed game,

- a strategy σ is winning whenever all its outcomes are winning;
- Cost of a winning strategy σ :

$$\text{cost}(\sigma) = \sup\{\text{cost}(\rho) \mid \rho \text{ outcome of } \sigma \text{ up to the target}\}$$

- Optimal cost:

$$\text{optcost}_{\mathcal{G}} = \inf_{\sigma \text{ winning strat.}} \text{cost}(\sigma)$$

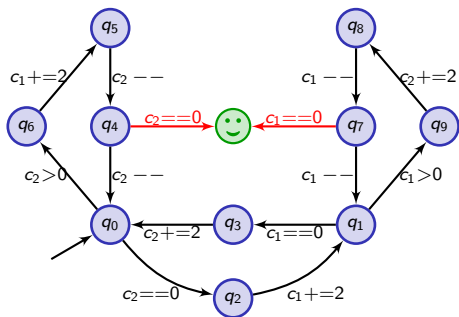
(set it to $+\infty$ if there is no winning strategy)

Two problems of interest

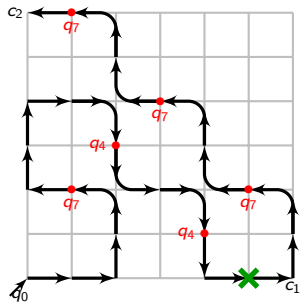
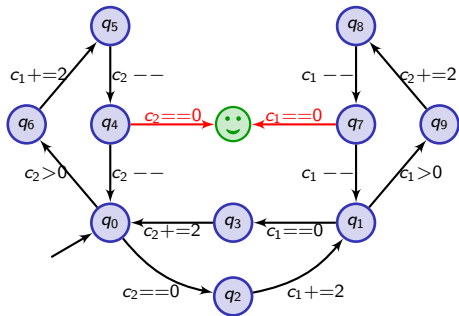
- The **value problem** asks, given \mathcal{G} and a threshold $\bowtie c$, whether $\text{optcost}_{\mathcal{G}} \bowtie c$?
- The **existence problem** asks, given \mathcal{G} and a threshold $\bowtie c$, whether there exists a winning strategy in \mathcal{G} such that $\text{cost}(\sigma) \bowtie c$?

Note: These problems are distinct...

The value of the game is 3, but no strategy has cost 3.



The value of the game is 3, but no strategy has cost 3.



- Weighted timed automata

In weighted timed automata, the optimal cost is an integer, and can be computed in PSPACE.

- Weighted timed automata

~~In weighted timed automata, the optimal cost is an integer, and can be computed in PSPACE.~~

The **value problem** is PSPACE-complete in weighted timed automata.
Almost-optimal winning schedules can be computed.

- Weighted timed automata

~~In weighted timed automata, the optimal cost is an integer, and can be computed in PSPACE.~~

The **value problem** is PSPACE-complete in weighted timed automata. Almost-optimal winning schedules can be computed.

- Weighted timed games

Turn-based optimal timed games are decidable in EXPTIME when automata have a single clock.

- Weighted timed automata

~~In weighted timed automata, the optimal cost is an integer, and can be computed in PSPACE.~~

The **value problem** is PSPACE-complete in weighted timed automata. Almost-optimal winning schedules can be computed.

- Weighted timed games

~~Turn-based optimal timed games are decidable in EXPTIME when automata have a single clock.~~

The **value problem** is decidable in EXPTIME in single-clock weighted timed games. Almost-optimal memoryless winning strategies can be computed.

- Weighted timed automata

~~In weighted timed automata, the optimal cost is an integer, and can be computed in PSPACE.~~

The **value problem** is PSPACE-complete in weighted timed automata. Almost-optimal winning schedules can be computed.

- Weighted timed games

~~Turn-based optimal timed games are decidable in EXPTIME when automata have a single clock.~~

The **value problem** is decidable in EXPTIME in single-clock weighted timed games. Almost-optimal memoryless winning strategies can be computed.

There is a symbolic algorithm to solve weighted timed games with a strongly non-Zeno cost.

- Weighted timed automata

~~In weighted timed automata, the optimal cost is an integer, and can be computed in PSPACE.~~

The **value problem** is PSPACE-complete in weighted timed automata. Almost-optimal winning schedules can be computed.

- Weighted timed games

~~Turn-based optimal timed games are decidable in EXPTIME when automata have a single clock.~~

The **value problem** is decidable in EXPTIME in single-clock weighted timed games. Almost-optimal memoryless winning strategies can be computed.

~~There is a symbolic algorithm to solve weighted timed games with a strongly non-Zeno cost.~~

The **value problem** can be decided in EXPTIME in weighted timed games with a strongly non-Zeno cost. Almost-optimal winning strategies can be computed.

- Weighted timed automata

~~In weighted timed automata, the optimal cost is an integer, and can be computed in PSPACE.~~

The **value problem** is PSPACE-complete in weighted timed automata. Almost-optimal winning schedules can be computed.

- Weighted timed games

~~Turn-based optimal timed games are decidable in EXPTIME when automata have a single clock.~~

The **value problem** is decidable in EXPTIME in single-clock weighted timed games. Almost-optimal memoryless winning strategies can be computed.

~~There is a symbolic algorithm to solve weighted timed games with a strongly non-Zeno cost.~~

The **value problem** can be decided in EXPTIME in weighted timed games with a strongly non-Zeno cost. Almost-optimal winning strategies can be computed.

In weighted timed games, the optimal cost cannot be computed, as soon as games have three clocks or more.

- Weighted timed automata

In weighted timed automata, the optimal cost is an integer, and can be computed in PSPACE.

The **value problem** is PSPACE-complete in weighted timed automata. Almost-optimal winning schedules can be computed.

- Weighted timed games

Turn-based optimal timed games are decidable in EXPTIME when automata have a single clock.

The **value problem** is decidable in EXPTIME in single-clock weighted timed games. Almost-optimal memoryless winning strategies can be computed.

There is a symbolic algorithm to solve weighted timed games with a strongly non-Zeno cost.

The **value problem** can be decided in EXPTIME in weighted timed games with a strongly non-Zeno cost. Almost-optimal winning strategies can be computed.

In weighted timed games, the optimal cost cannot be computed, as soon as games have three clocks or more.

The **existence problem** is undecidable in weighted timed games.

Our recent developments

- 1 The **value problem** is undecidable in weighted timed games
 - ~> Intellectually satisfactory to not have this discrepancy in the set of results
 - ~> Proof based on a diagonal construction (originally proposed in the context of quantitative temporal logics [BMM14])

Our recent developments

- 1 The **value problem** is undecidable in weighted timed games
 - ~> Intellectually satisfactory to not have this discrepancy in the set of results
 - ~> Proof based on a diagonal construction (originally proposed in the context of quantitative temporal logics [BMM14])
- 2 An **approximation algorithm** for a large class of weighted timed games (that comprises the class of games used for proving the above undecidability)
 - Almost-optimality in practice should be sufficient
 - Even when we know how to compute the value, we are only able to synthesize almost-optimal strategies...

Optimal cost is computable...

... when cost is strongly non-zero.

[AM04,BCFL04]

That is, there exists $\kappa > 0$ such that for every region cycle C , for every real run ρ read on C ,

$$\text{cost}(\rho) \geq \kappa$$

Optimal cost is not computable...

... when cost is almost-strongly non-zero.

That is, there exists $\kappa > 0$ such that for every region cycle C , for every real run ρ read on C ,

$$\text{cost}(\rho) \geq \kappa \quad \text{or} \quad \text{cost}(\rho) = 0$$

Note: In both cases, we can assume $\kappa = 1$.

Optimal cost is computable...

... when cost is strongly non-zero.

[AM04,BCFL04]

That is, there exists $\kappa > 0$ such that for every region cycle C , for every real run ρ read on C ,

$$\text{cost}(\rho) \geq \kappa$$

Optimal cost is not computable... but is approximable!

... when cost is almost-strongly non-zero.

[BJM15]

That is, there exists $\kappa > 0$ such that for every region cycle C , for every real run ρ read on C ,

$$\text{cost}(\rho) \geq \kappa \quad \text{or} \quad \text{cost}(\rho) = 0$$

Note: In both cases, we can assume $\kappa = 1$.

Approximation of the optimal cost

Theorem

Let \mathcal{G} be a weighted timed game, in which the cost is almost-strongly non-zero. For every $\epsilon > 0$, one can compute:

- two values v_ϵ^- and v_ϵ^+ such that

$$|v_\epsilon^+ - v_\epsilon^-| < \epsilon \quad \text{and} \quad v_\epsilon^- \leq \text{optcost}_{\mathcal{G}} \leq v_\epsilon^+$$

Approximation of the optimal cost

Theorem

Let \mathcal{G} be a weighted timed game, in which the cost is almost-strongly non-zero. For every $\epsilon > 0$, one can compute:

- two values v_ϵ^- and v_ϵ^+ such that

$$|v_\epsilon^+ - v_\epsilon^-| < \epsilon \quad \text{and} \quad v_\epsilon^- \leq \text{optcost}_{\mathcal{G}} \leq v_\epsilon^+$$

- one strategy σ_ϵ such that

$$\text{optcost}_{\mathcal{G}} \leq \text{cost}(\sigma_\epsilon) \leq \text{optcost}_{\mathcal{G}} + \epsilon$$

It is an ϵ -optimal winning strategy.

Approximation of the optimal cost

Theorem

Let \mathcal{G} be a weighted timed game, in which the cost is almost-strongly non-zero. For every $\epsilon > 0$, one can compute:

- two values v_ϵ^- and v_ϵ^+ such that

$$|v_\epsilon^+ - v_\epsilon^-| < \epsilon \quad \text{and} \quad v_\epsilon^- \leq \text{optcost}_{\mathcal{G}} \leq v_\epsilon^+$$

- one strategy σ_ϵ such that

$$\text{optcost}_{\mathcal{G}} \leq \text{cost}(\sigma_\epsilon) \leq \text{optcost}_{\mathcal{G}} + \epsilon$$

It is an ϵ -optimal winning strategy.

- Standard technics: unfold the game to get more precision, and compute two adjacency sequences

Approximation of the optimal cost

Theorem

Let \mathcal{G} be a weighted timed game, in which the cost is almost-strongly non-zero. For every $\epsilon > 0$, one can compute:

- two values v_ϵ^- and v_ϵ^+ such that

$$|v_\epsilon^+ - v_\epsilon^-| < \epsilon \quad \text{and} \quad v_\epsilon^- \leq \text{optcost}_{\mathcal{G}} \leq v_\epsilon^+$$

- one strategy σ_ϵ such that

$$\text{optcost}_{\mathcal{G}} \leq \text{cost}(\sigma_\epsilon) \leq \text{optcost}_{\mathcal{G}} + \epsilon$$

It is an ϵ -optimal winning strategy.

- Standard technics: unfold the game to get more precision, and compute two adjacency sequences
- ↪ This is not possible here
There might be runs with prefixes of arbitrary length and cost 0 (e.g. the game of the undecidability proof)

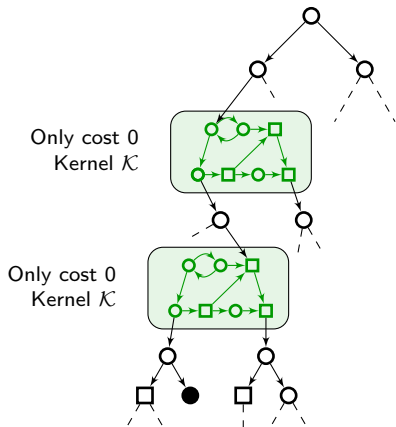
Idea for approximation

Idea

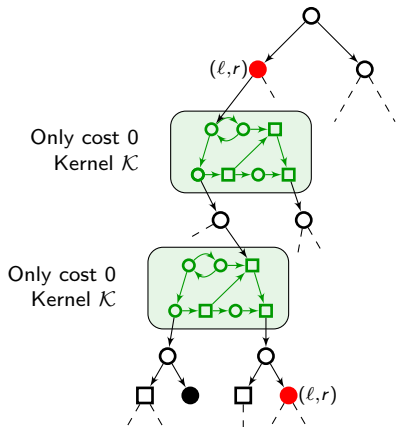
Only partially unfold the game:

- Keep components with cost 0 untouched – we call it the **kernel**
- Unfold the rest of the game

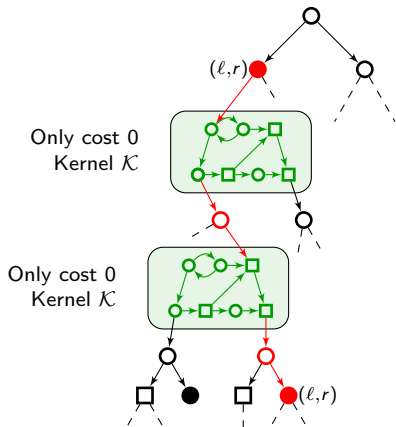
Semi-unfolding



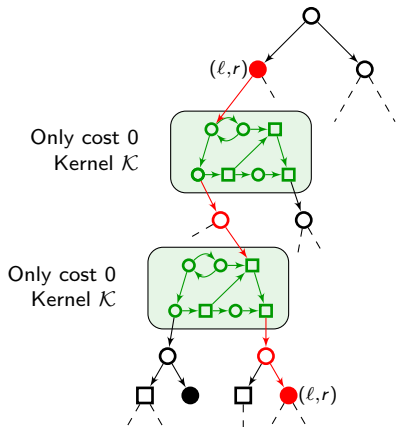
Semi-unfolding



Semi-unfolding

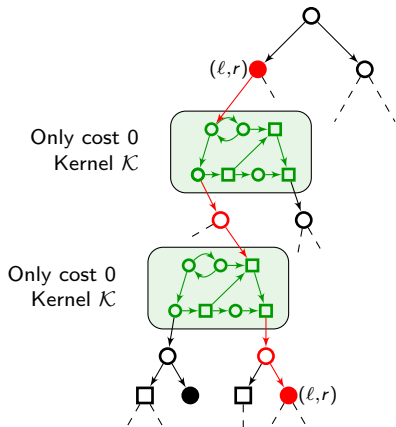


Semi-unfolding



Hypothesis:
 $\text{cost} > 0$ implies $\text{cost} \geq \kappa$

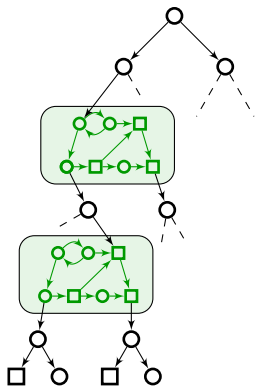
Semi-unfolding



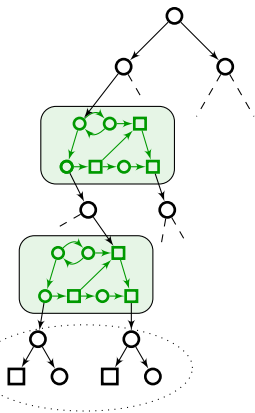
Hypothesis:
 $\text{cost} > 0$ implies $\text{cost} \geq \kappa$

Conclusion: we can stop unfolding the game after N steps
(e.g. $N = (M + 2) \cdot |\mathcal{R}(\mathcal{A})|$, where M is a pre-computed bound on $\text{optcost}_{\mathcal{G}}$)

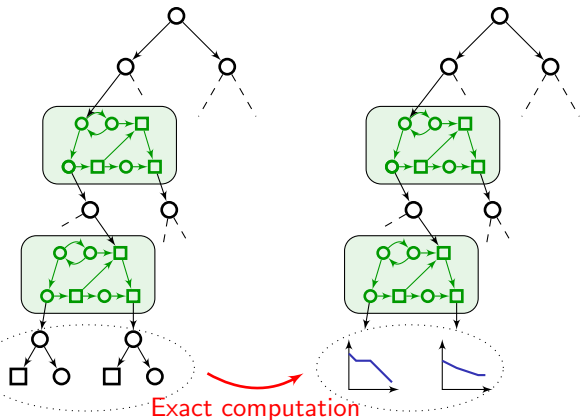
Approximation scheme



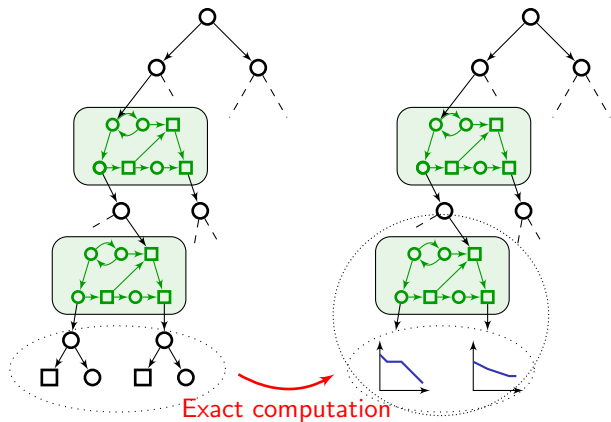
Approximation scheme



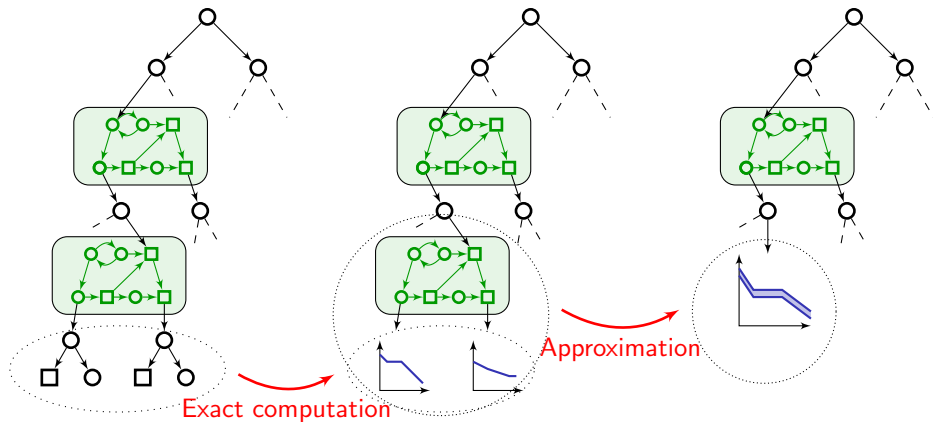
Approximation scheme



Approximation scheme



Approximation scheme

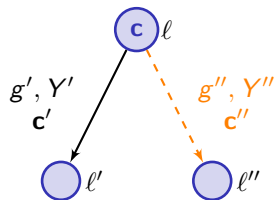


First step: Tree-like parts

↪ Goes back to [LMM02]

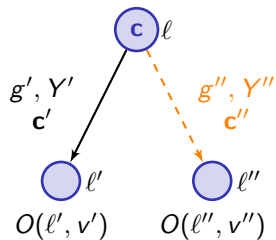
First step: Tree-like parts

\rightsquigarrow Goes back to [LMM02]



First step: Tree-like parts

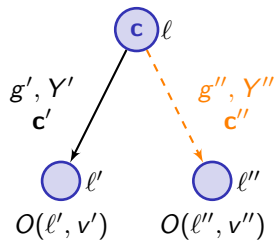
\rightsquigarrow Goes back to [LMM02]



$$O(l, v) =$$

First step: Tree-like parts

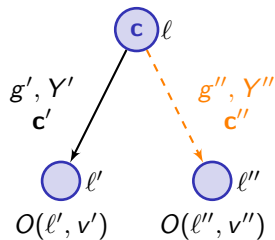
\rightsquigarrow Goes back to [LMM02]



$$O(l, v) = \inf_{t' | v+t' \models g'}$$

First step: Tree-like parts

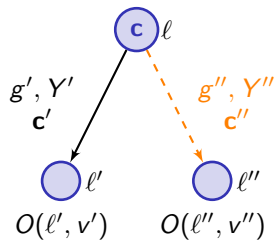
\rightsquigarrow Goes back to [LMM02]



$$O(l, v) = \inf_{t' | v+t' \models g'} \max(\quad , \quad)$$

First step: Tree-like parts

↷ Goes back to [LMM02]



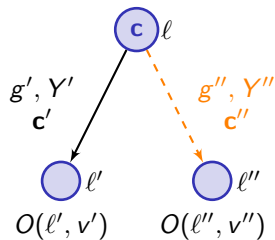
$$O(\ell, v) = \inf_{t' | v+t' \models g'} \max((\alpha), \quad)$$

$$(\alpha) = t'c + c' + O(\ell', v')$$

$$v' = [Y' \leftarrow 0](v+t')$$

First step: Tree-like parts

↷ Goes back to [LMM02]



$$O(\ell, v) = \inf_{t' | v+t' \models g'} \max((\alpha), (\beta))$$

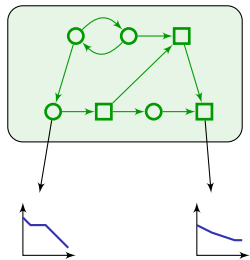
$$(\alpha) = t'c + c' + O(\ell', v')$$

$$(\beta) = \sup_{t'' \leq t' | v+t'' \models g''} t''c + c'' + O(\ell'', v'')$$

$$v' = [Y' \leftarrow 0](v+t')$$

$$v'' = [Y'' \leftarrow 0](v+t'')$$

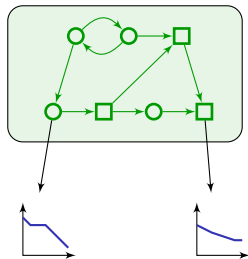
Second step: Kernels



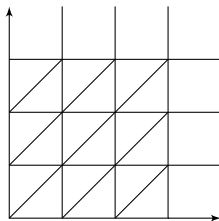
Output cost functions f

Second step: Kernels

- 1 Refine the regions such that f differs of at most ϵ within a small region

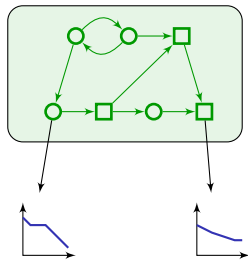


Output cost functions f

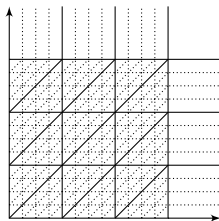


Second step: Kernels

- 1 Refine the regions such that f differs of at most ϵ within a small region

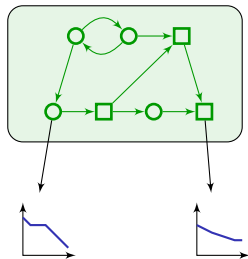


Output cost functions f

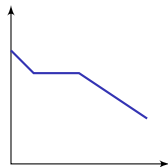
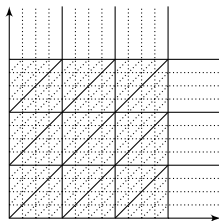


Second step: Kernels

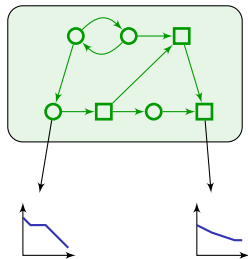
- 1 Refine the regions such that f differs of at most ϵ within a small region



Output cost functions f

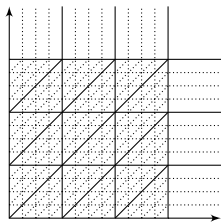


Second step: Kernels

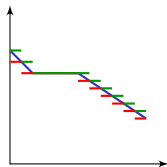


Output cost functions f

- 1 Refine the regions such that f differs of at most ϵ within a small region

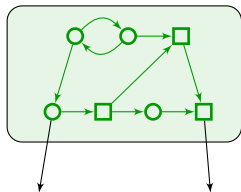


- 2 Under- and over-approximate by piecewise constant functions f_ϵ^- and f_ϵ^+



Second step: Kernels

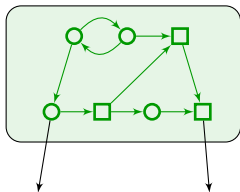
- 3 Refine/split the kernel along the new small regions and fix f_ϵ^- or f_ϵ^+ , write f_ϵ



f_ϵ^- : constant f_ϵ^+ : constant

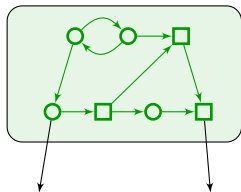
Second step: Kernels

- 3 Refine/split the kernel along the new small regions and fix f_ϵ^- or f_ϵ^+ , write f_ϵ
- 4 Since cost is 0 everywhere, the resulting game is nothing more than a **reachability timed game** with an order on target (output) edges (given by f_ϵ)



f_ϵ^- : constant f_ϵ^+ : constant

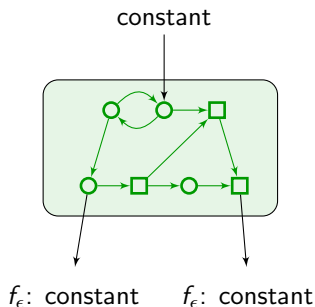
Second step: Kernels



f_ϵ^- : constant f_ϵ^+ : constant

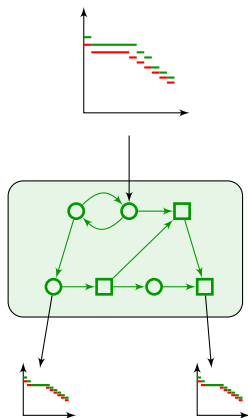
- 3 Refine/split the kernel along the new small regions and fix f_ϵ^- or f_ϵ^+ , write f_ϵ
- 4 Since cost is 0 everywhere, the resulting game is nothing more than a **reachability timed game** with an order on target (output) edges (given by f_ϵ)
- 5 Those can be solved using standard technics based on attractors: small regions are sufficient, and the local optimal cost (for output f_ϵ) is constant within a small region

Second step: Kernels



- 3 Refine/split the kernel along the new small regions and fix f_ϵ^- or f_ϵ^+ , write f_ϵ
- 4 Since cost is 0 everywhere, the resulting game is nothing more than a **reachability timed game** with an order on target (output) edges (given by f_ϵ)
- 5 Those can be solved using standard technics based on attractors: small regions are sufficient, and the local optimal cost (for output f_ϵ) is constant within a small region

Second step: Kernels



- 3 Refine/split the kernel along the new small regions and fix f_{ϵ}^{-} or f_{ϵ}^{+} , write f_{ϵ}
 - 4 Since cost is 0 everywhere, the resulting game is nothing more than a **reachability timed game** with an order on target (output) edges (given by f_{ϵ})
 - 5 Those can be solved using standard technics based on attractors: small regions are sufficient, and the local optimal cost (for output f_{ϵ}) is constant within a small region
- ~ We have computed ϵ -approximations of the optimal cost, which are constant within small regions. Corresponding strategies can be inferred

Conclusion

Summary of the talk

- Very quick overview of results concerning the optimal reachability problem in weighted timed games
- Some new insight into the value problem for this model:
 - Undecidability of this problem
 - Approximability of the optimal cost (under some conditions)

Conclusion

Summary of the talk

- Very quick overview of results concerning the optimal reachability problem in weighted timed games
- Some new insight into the value problem for this model:
 - Undecidability of this problem
 - Approximability of the optimal cost (under some conditions)

Future work

- Improve the approximation scheme ($2\text{EXP}(|\mathcal{G}|) \cdot (1/\epsilon)^{|\mathcal{X}|}$)
- Extend to the whole class of weighted timed games, or understand why it is not possible
- Assume stochastic uncertainty?
- Multiplayer setting?