

# Energy games

Nicolas Markey (CNRS)  
IRISA, team SUMO

Based on joint works with

Patricia Bouyer

Uli Fahrenberg

Piotr Hofman

Kim G. Larsen

Simon Laursen

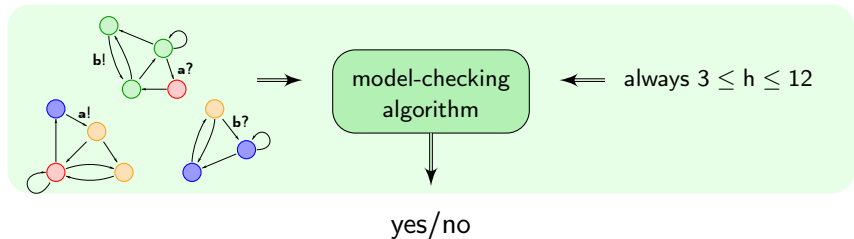
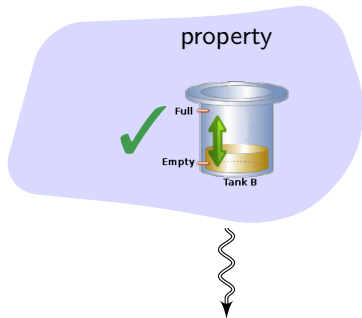
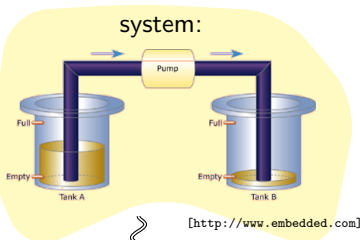
Mickael Randour

Jiri Srba

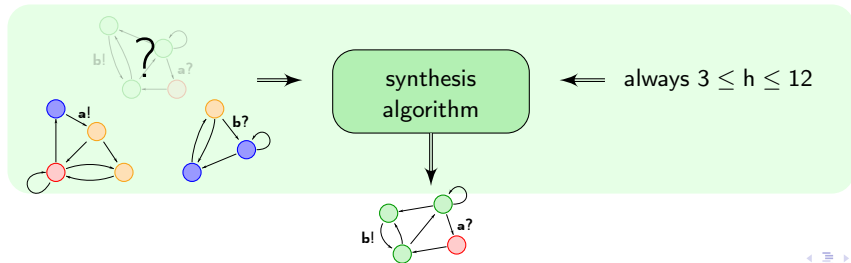
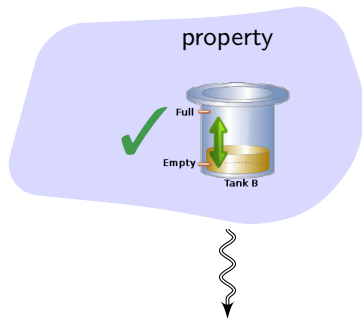
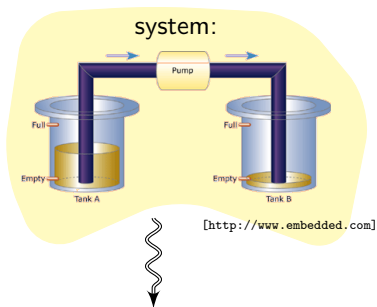
Martin Zimmermann

June 29, 2017

# Model checking and synthesis

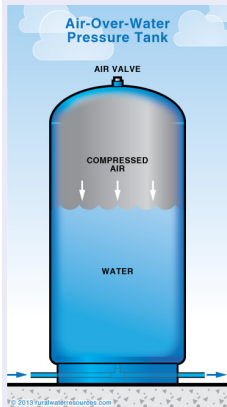


# Model checking and synthesis



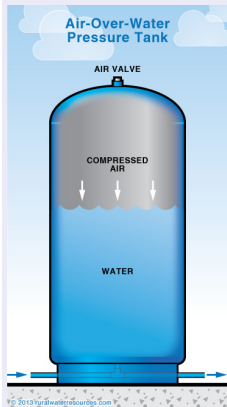
# (Average-)energy objectives: an example

## Pressure-tank case study [CJL<sup>+</sup>09]



# (Average-)energy objectives: an example

## Pressure-tank case study [CJL<sup>+</sup>09]

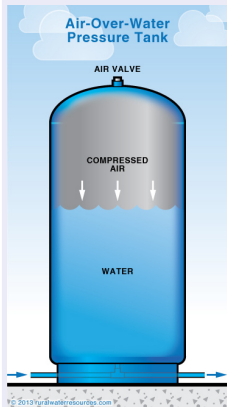


### Objectives:

- keep water level within given bounds
- minimize average level

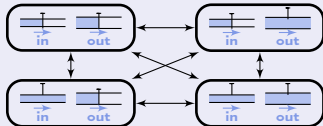
# (Average-)energy objectives: an example

## Pressure-tank case study [CJL<sup>+</sup>09]



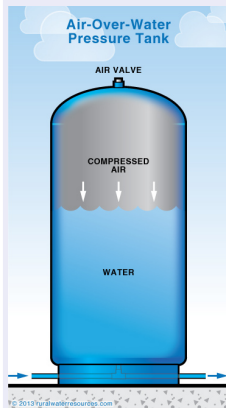
### Objectives:

- keep water level within given bounds
- minimize average level



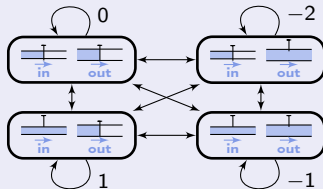
# (Average-)energy objectives: an example

## Pressure-tank case study [CJL<sup>+</sup>09]



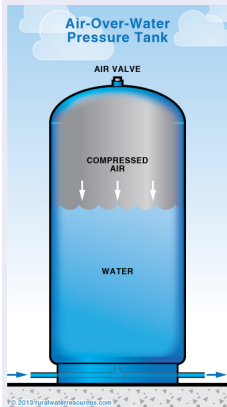
### Objectives:

- keep water level within given bounds
- minimize average level



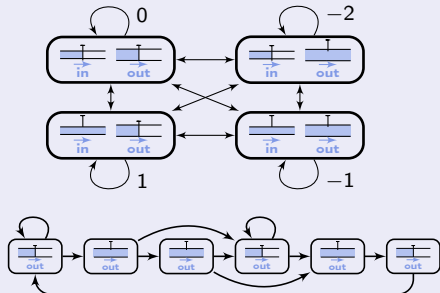
# (Average-)energy objectives: an example

## Pressure-tank case study [CJL<sup>+</sup>09]



### Objectives:

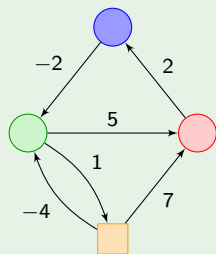
- keep water level within given bounds
- minimize average level





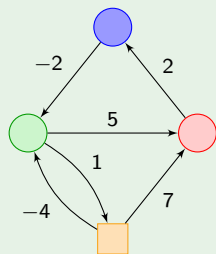
# Games on weighted graphs

## Example



# Games on weighted graphs

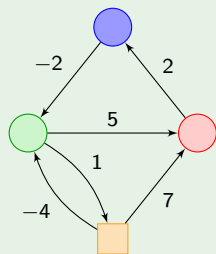
## Example



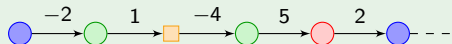
- states:  $\mathcal{S} = \mathcal{S}_{\bullet} \uplus \mathcal{S}_{\square}$
- weighted transitions:  $T \subseteq \mathcal{S} \times \mathbb{Z} \times \mathcal{S}$

# Games on weighted graphs

## Example

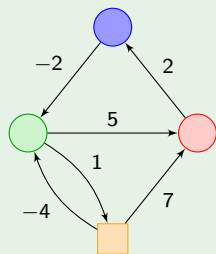


- **states:**  $\mathcal{S} = \mathcal{S}_{\bullet} \uplus \mathcal{S}_{\square}$
- **weighted transitions:**  $T \subseteq \mathcal{S} \times \mathbb{Z} \times \mathcal{S}$
- **run:** sequence of consecutive transitions:

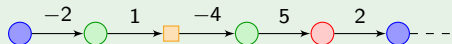


# Games on weighted graphs

## Example



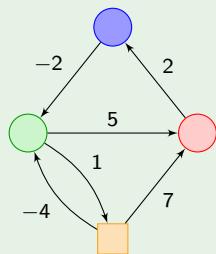
- **states:**  $\mathcal{S} = \mathcal{S}_{\bullet} \uplus \mathcal{S}_{\square}$
- **weighted transitions:**  $T \subseteq \mathcal{S} \times \mathbb{Z} \times \mathcal{S}$
- **run:** sequence of consecutive transitions:



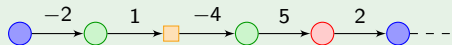
- **strategy:** transition to take depending on state/history:

# Games on weighted graphs

## Example



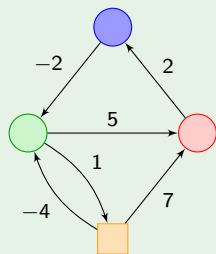
- **states:**  $\mathcal{S} = \mathcal{S}_{\bullet} \uplus \mathcal{S}_{\square}$
- **weighted transitions:**  $T \subseteq \mathcal{S} \times \mathbb{Z} \times \mathcal{S}$
- **run:** sequence of consecutive transitions:



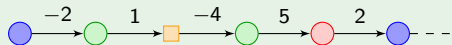
- **strategy:** transition to take depending on state/history:  
 $\sigma_{\bullet}$ : always go to  $\square$  (from  $\circ$ )

# Games on weighted graphs

## Example



- **states:**  $S = S_{\bullet} \uplus S_{\square}$
- **weighted transitions:**  $T \subseteq S \times \mathbb{Z} \times S$
- **run:** sequence of consecutive transitions:



- **strategy:** transition to take depending on state/history:
  - $\sigma_{\bullet}$ : always go to  $\square$  (from  $\circ$ )
  - $\sigma'_{\bullet}$ : alternate between  $\circ$  and  $\square$  (from  $\circ$ )

# Quantitative objectives

## Decision problems

- **shortest path:**

is it possible to go from **source** to **target** with accumulated weight less than a given **threshold**?

# Quantitative objectives

## Decision problems

- **shortest path:**

is it possible to go from **source** to **target** with accumulated weight less than a given **threshold**?

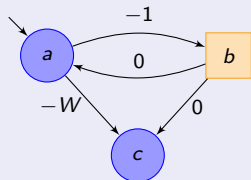
~ in **polynomial time** for 1-player games

(e.g. Bellman-Ford algorithm)

~ in **pseudo-polynomial time** for 2-player games [BGHM17]

in  $NP \cap coNP$ , PTIME-hard

**PTIME-complete** with **nonnegative** weights [KBB<sup>+</sup>08]





# Quantitative objectives

## Decision problems

- **mean-payoff objective:**  
from a given **source**, is it possible to make the average weight below a given **threshold** (in the long run)?

# Quantitative objectives

## Decision problems

- **mean-payoff objective:**

from a given **source**, is it possible to make the average weight below a given **threshold** (**in the long run**)?

~> in **polynomial time** for 1-player games

(e.g. Karp algorithm [Kar78])

~> in **pseudo-polynomial time** for 2-player games

in **NP**  $\cap$  **coNP**, PTIME-hard

[ZP96]

[Kar78] Karp. A characterization of the minimum cycle mean in a digraph. Discr.Math., 1978.

[ZP96] Zwick, Paterson. The complexity of mean payoff games on graphs. TCS, 1996.

## Energy objectives

### Energy level

- energy level:  $EL(\pi_{\leq n}) = \sum_{i \leq n} w(s_i \rightarrow s_{i+1})$  [aka. total payoff]

# Energy objectives

## Energy level

- **energy level**:  $EL(\pi_{\leq n}) = \sum_{i \leq n} w(s_i \rightarrow s_{i+1})$  [aka. **total payoff**]

## Quantitative objectives

- **shortest path**: minimize energy level when reaching a target
- **mean-payoff**: minimize ratio (energy level/path length) in the long run

# Energy objectives

## Energy level

- **energy level**:  $EL(\pi_{\leq n}) = \sum_{i \leq n} w(s_i \rightarrow s_{i+1})$  [aka. **total payoff**]

## Quantitative objectives

- **shortest path**: minimize energy level when reaching a target
- **mean-payoff**: minimize ratio (energy level/path length) in the long run
- **energy objectives**: maintain energy level within some bounds
  - above lower bound  $L$
  - between bounds  $L$  and  $U$

# Solving lower-bounded energy games

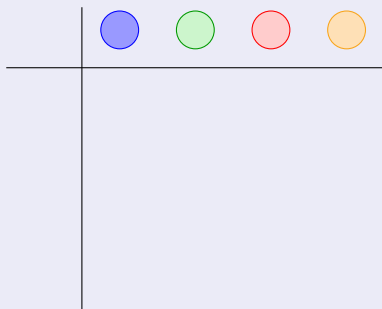
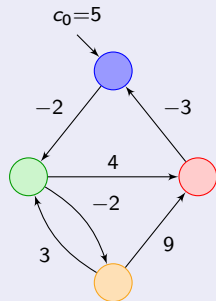
## 1-player case [BFLMS08]

- aim: maintain energy level above  $L \rightsquigarrow$  maximize energy level
- Bellman-Ford-like algorithm to compute maximal remaining energy after  $k$  steps

# Solving lower-bounded energy games

## 1-player case [BFLMS08]

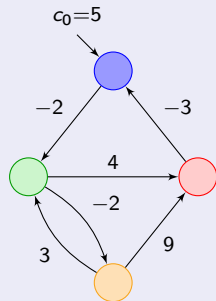
- aim: maintain energy level above  $L \rightsquigarrow$  maximize energy level
- Bellman-Ford-like algorithm to compute maximal remaining energy after  $k$  steps







# Solving lower-bounded energy games

## 1-player case [BFLMS08]

- aim: maintain energy level above  $L \rightsquigarrow$  maximize energy level
- Bellman-Ford-like algorithm to compute maximal remaining energy after  $k$  steps



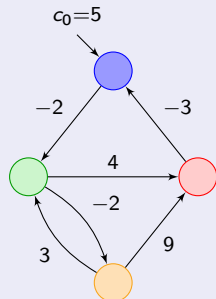
				
0	5	$-\infty$	$-\infty$	$-\infty$







# Solving lower-bounded energy games

## 1-player case [BFLMS08]

- aim: maintain energy level above  $L \rightsquigarrow$  maximize energy level
- Bellman-Ford-like algorithm to compute maximal remaining energy after  $k$  steps

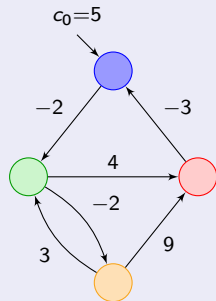


				
0	5	$-\infty$	$-\infty$	$-\infty$
1	5	3	$-\infty$	$-\infty$

# Solving lower-bounded energy games

## 1-player case [BFLMS08]

- aim: maintain energy level above  $L \rightsquigarrow$  maximize energy level
- Bellman-Ford-like algorithm to compute maximal remaining energy after  $k$  steps

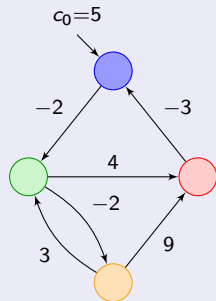






0	5	$-\infty$	$-\infty$	$-\infty$
1	5	3	$-\infty$	$-\infty$
2	5	3	7	1

# Solving lower-bounded energy games

## 1-player case [BFLMS08]

- aim: maintain energy level above  $L \rightsquigarrow$  maximize energy level
- Bellman-Ford-like algorithm to compute maximal remaining energy after  $k$  steps

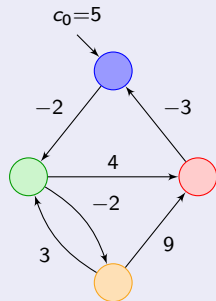






				
0	5	$-\infty$	$-\infty$	$-\infty$
1	5	3	$-\infty$	$-\infty$
2	5	3	7	1
3	5	4	10	1

# Solving lower-bounded energy games

## 1-player case [BFLMS08]

- aim: maintain energy level above  $L \rightsquigarrow$  maximize energy level
- Bellman-Ford-like algorithm to compute maximal remaining energy after  $k$  steps



				
0	5	$-\infty$	$-\infty$	$-\infty$
1	5	3	$-\infty$	$-\infty$
2	5	3	7	1
3	5	4	10	1
4	7	4	10	2

# Solving lower-bounded energy games

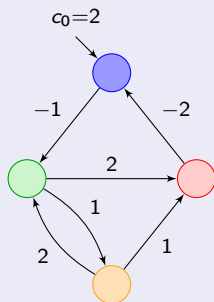
## 2-player case

- aim: **maximize** (resp. **minimize**) energy level
- both players have **memoryless optimal strategies**
- deciding the winner is in  **$NP \cap coNP$**
- **mean-payoff games** are logspace-reducible to  $L$ -energy games

# Solving interval-bounded energy games

## Reduction to safety condition in pseudo-polynomial graph

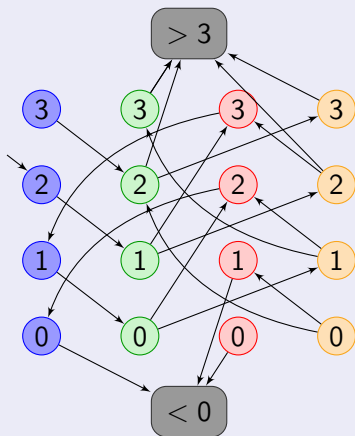
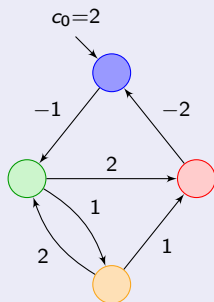
Objective: keep energy level  
between  $L = 0$  and  $U = 3$



# Solving interval-bounded energy games

## Reduction to safety condition in pseudo-polynomial graph

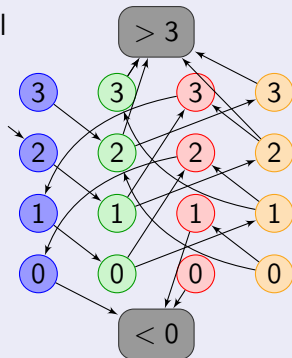
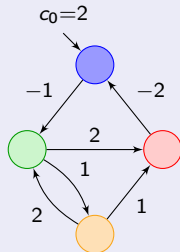
Objective: keep energy level  
between  $L = 0$  and  $U = 3$



# Solving interval-bounded energy games

## Reduction to safety condition in pseudo-polynomial graph

Objective: keep energy level  
between  $L = 0$  and  $U = 3$



## Theorem ([BFLMS08,FJ13])

*1-player interval-bounded energy games are PSPACE-complete.*

*2-player interval-bounded energy games are EXPTIME-complete.*

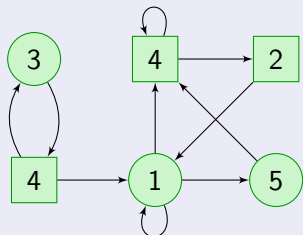
[BFLMS08] Bouyer *et al.* Infinite Runs in Weighted Timed Automata [...] FORMATS, 2008.

[FJ13] Fearnley, Jurdziński. Reachability in two-clock timed automata is PSPACE-complete. ICALP, 2013.



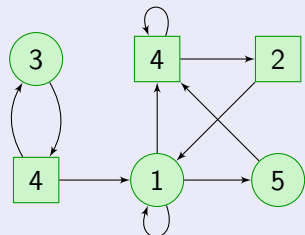
# Energy parity games

## Parity games



# Energy parity games

## Parity games



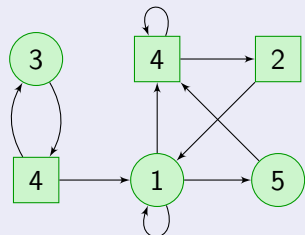
$M(\gamma)$  = maximal value seen along  $\gamma$ .

Objective of Player 1:

make  $M(\gamma)$  even for any outcome.

# Energy parity games

## Parity games



$M(\gamma)$  = maximal value seen along  $\gamma$ .

Objective of Player 1:

make  $M(\gamma)$  even for any outcome.

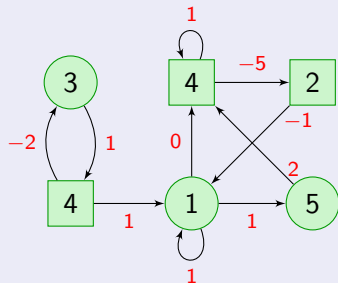
## Theorem

*Both players have memoryless optimal strategies.*

*Deciding the winner is in  $\text{NP} \cap \text{coNP}$ .*

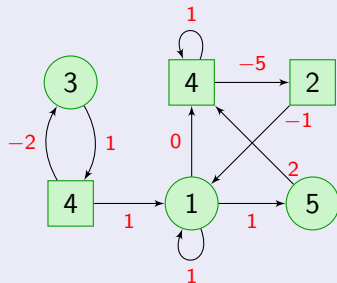
# Energy parity games

## Energy parity games



# Energy parity games

## Energy parity games

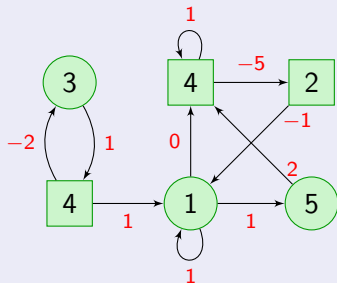


Objective of Player 1: for any  $\gamma$

- make  $M(\gamma)$  even
- keep energy above  $L$  all along  $\gamma$

# Energy parity games

## Energy parity games



Objective of Player 1: for any  $\gamma$

- make  $M(\gamma)$  even
- keep energy above  $L$  all along  $\gamma$

## Theorem ([CD12])

*Player 2 has memoryless optimal strategies.*

*Player 1 has optimal strategies combining 2 memoryless strategies.*

*Deciding the winner is in  $\text{NP} \cap \text{coNP}$ .*

## Average-energy games

Average energy is **not** mean-payoff

- **mean payoff** = average of weight on transitions

$$MP(\pi_{\leq n}) = \limsup_{n \rightarrow \infty} \frac{1}{n} EL(\pi_{\leq n})$$

# Average-energy games

## Average energy is **not** mean-payoff

- **mean payoff** = average of weight on transitions

$$MP(\pi_{\leq n}) = \limsup_{n \rightarrow \infty} \frac{1}{n} EL(\pi_{\leq n})$$

- **average energy** = average of accumulated weight

$$AE(\pi_{\leq n}) = \limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n EL(\pi_{\leq i})$$



# Average-energy games

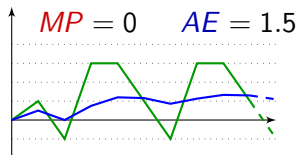
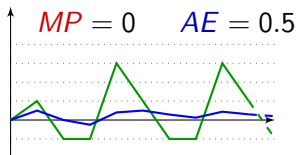
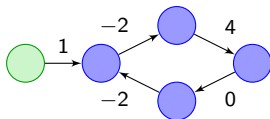
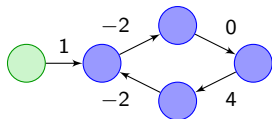
Average energy is **not** mean-payoff

- **mean payoff** = average of weight on transitions

$$MP(\pi_{\leq n}) = \limsup_{n \rightarrow \infty} \frac{1}{n} EL(\pi_{\leq n})$$

- **average energy** = average of accumulated weight

$$AE(\pi_{\leq n}) = \limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{i=1}^n EL(\pi_{\leq i})$$



# Average-energy games

## Theorem ([BMRL15])

- *1-player AE games can be solved in PTIME (memoryless strategies are sufficient)*

# Average-energy games

## Theorem ([BMRL15])

- 1-player AE games can be solved in **PTIME** (memoryless strategies are sufficient)
  - ▶ compute  $\overline{AE}(C_{k,s})$  for all simple cycles of length  $k$  on  $s$ ;
  - minimize  $EL(\rho_{s_0 \rightarrow s}) + \overline{AE}(C_s)$ .

# Average-energy games

## Theorem ([BMRL15])

- 1-player AE games can be solved in PTIME (memoryless strategies are sufficient)
  - compute  $\overline{AE}(C_{k,s})$  for all simple cycles of length  $k$  on  $s$ ;
  - ▶ minimize  $EL(\rho_{s_0 \rightarrow s}) + \overline{AE}(C_s)$ .

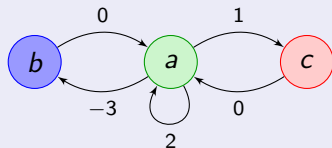
# Average-energy games

## Theorem ([BMRL15])

- 1-player AE games can be solved in **PTIME** (memoryless strategies are sufficient)
  - compute  $\overline{AE}(C_{k,s})$  for all simple cycles of length  $k$  on  $s$ ;
  - minimize  $EL(\rho_{s_0 \rightarrow s}) + \overline{AE}(C_s)$ .
- in 2-player AE games, both players have **memoryless optimal strategies**
- deciding the winner is in  **$NP \cap coNP$**
- **mean-payoff games** are logspace-reducible to AE games

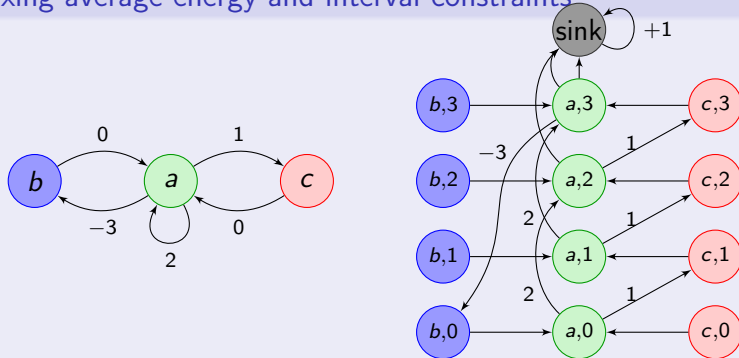
# Energy- and average-energy constraints

## Mixing average-energy and interval constraints



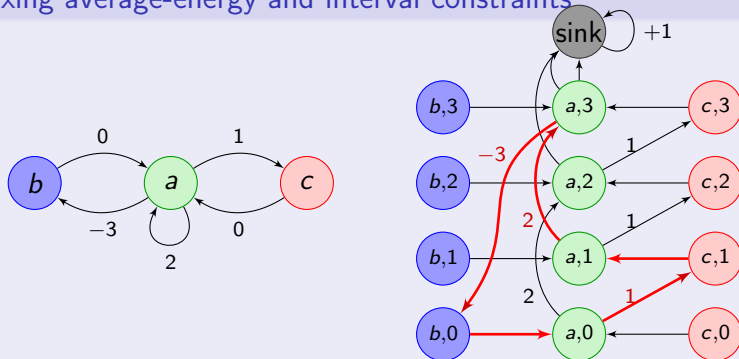
# Energy- and average-energy constraints

## Mixing average-energy and interval constraints



# Energy- and average-energy constraints

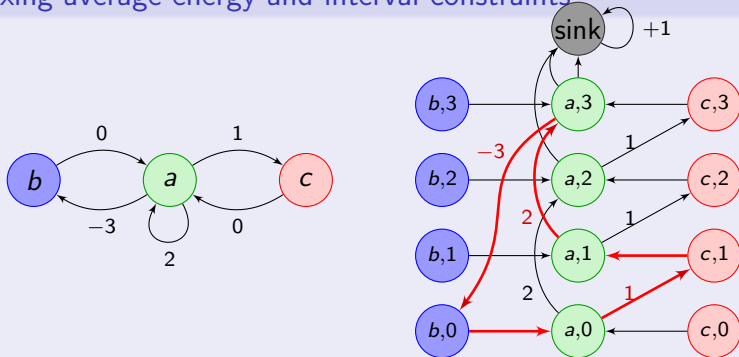
## Mixing average-energy and interval constraints





# Energy- and average-energy constraints

## Mixing average-energy and interval constraints

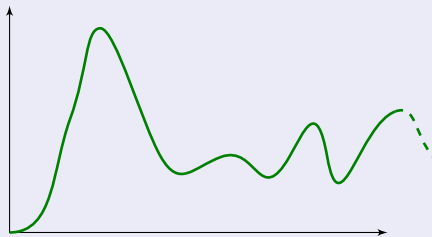


## Theorem ([BMRL15])

1-player *AELU*-games are in **EXPTIME**, and **PSPACE**-hard.  
2-player *AELU*-games are **EXPTIME**-complete.

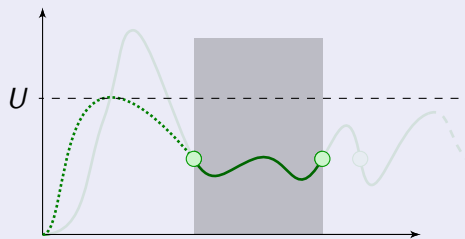
# Energy- and average-energy constraints

## Mixing average-energy and lower-bound constraints



# Energy- and average-energy constraints

## Mixing average-energy and lower-bound constraints

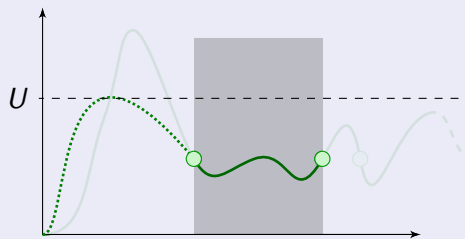


Bound peak height  $U$ :

- pseudo-polynomial for 1-player games
- 2-exponential for 2-player games

# Energy- and average-energy constraints

## Mixing average-energy and lower-bound constraints



Bound peak height  $U$ :

- pseudo-polynomial for 1-player games
- 2-exponential for 2-player games

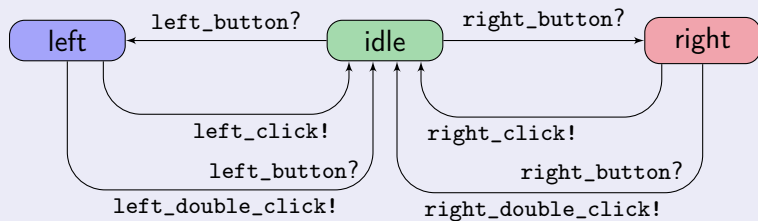
## Theorem ([BHM<sup>+</sup>17])

1-player AEL-games are in EXPTIME, and NP-hard.

2-player AEL-games are in 2-EXPTIME, and EXPSPACE-hard.

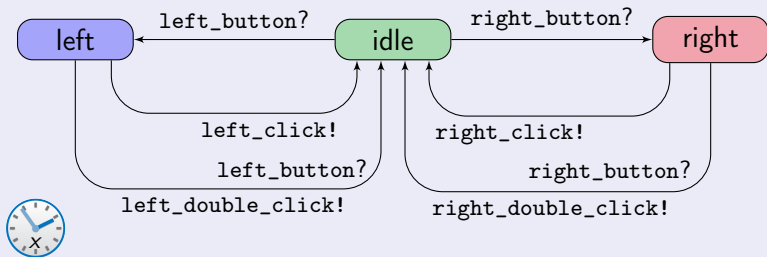
# Timed energy games

## Timed automata: example of a computer mouse



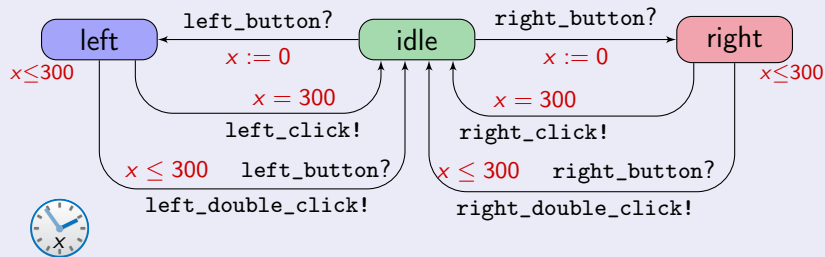
# Timed energy games

## Timed automata: example of a computer mouse



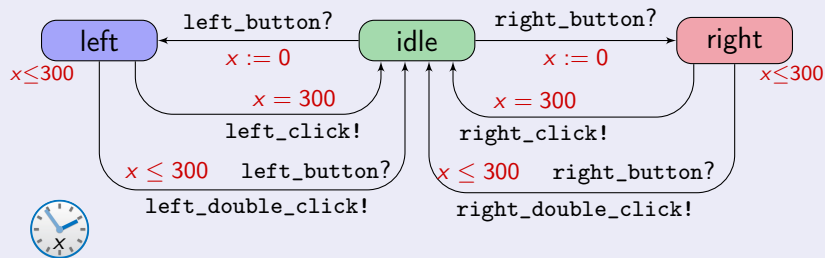
# Timed energy games

## Timed automata: example of a computer mouse



# Timed energy games

## Timed automata: example of a computer mouse



## Theorem ([AD94,AMP<sup>+</sup>98])

Reachability in timed automata is **PSPACE-complete**.

Reachability in timed games is **EXPTIME-complete**.

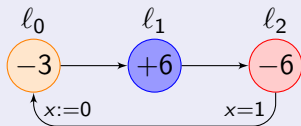
[AD94] Alur and Dill. A Theory of Timed Automata. TCS, 1994.

[AMP<sup>+</sup>98] Asarin et al.. Controller Synthesis for Timed Automata. SSC, 1998.



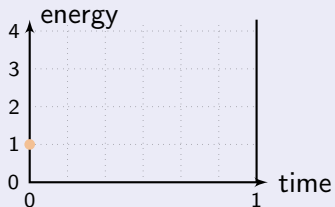
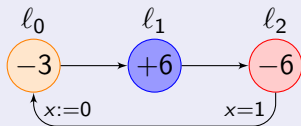
# Timed energy games

## Weighted timed automata



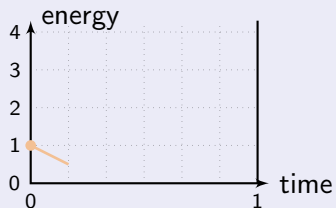
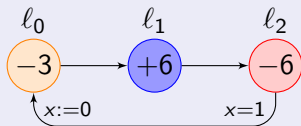
# Timed energy games

## Weighted timed automata



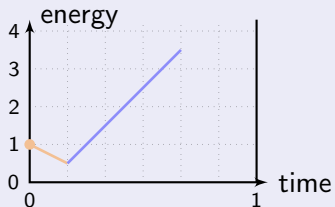
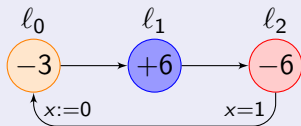
# Timed energy games

## Weighted timed automata



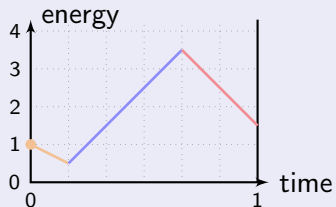
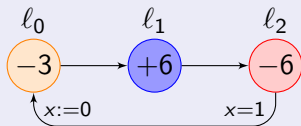
# Timed energy games

## Weighted timed automata



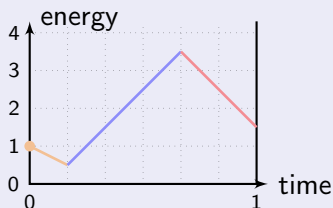
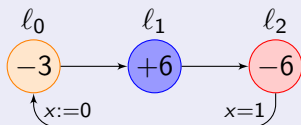
# Timed energy games

## Weighted timed automata



# Timed energy games

## Weighted timed automata



## Theorem ([BFLMS08,BFLM10])

- For 1 player:
  - lower-bound problem for 1-clock timed automata is in **EXPTIME**
  - interval problem for 2-clock timed automata is *undecidable*
- For 2 players:
  - interval problem for 1-clock timed automata is *undecidable*

[BFLMS08] Bouyer *et al.* Infinite Runs in Weighted Timed Automata [...] FORMATS, 2008.

[BFLM10] Bouyer *et al.* Timed Automata with Observers under Energy Constraints. HSCC, 2010.

# Conclusion and future works

## Conclusion

- **Weighted games**, in particular **energy games**, conveniently model resource-management problems;
- they are rather **well-understood**, but with significant open problems.
- **no real tools available**, only prototypes.

## Future works

- extend to **stochastic strategies, stochastic games**
- **multiple-player** quantitative games
- combine weighted timed games **with imprecisions**