

Looking at Mean-Payoff through Foggy Windows

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Université Libre de Bruxelles

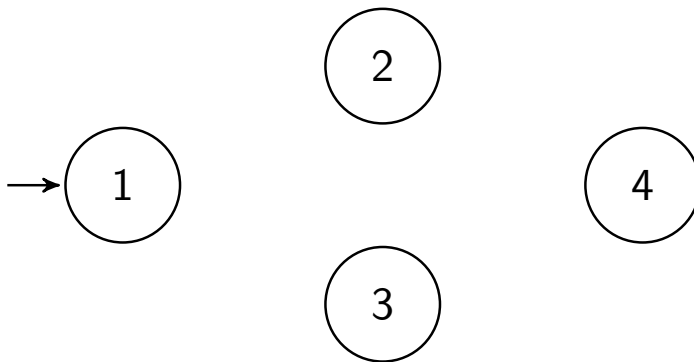
Cassting meeting (Oct'15) @ Cachan

Outline

- 1 Mean-payoff games with PO
- 2 Window objectives
 - Defs. and nice properties
 - Partial-Observation generalizations
- 3 Results
 - Lower bounds
 - Upper bounds
- 4 Conclusions

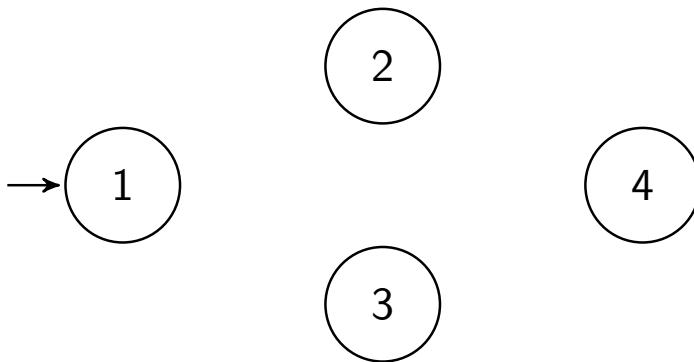
MPGs with partial-observation: example

- (Q, q_I, \dots)



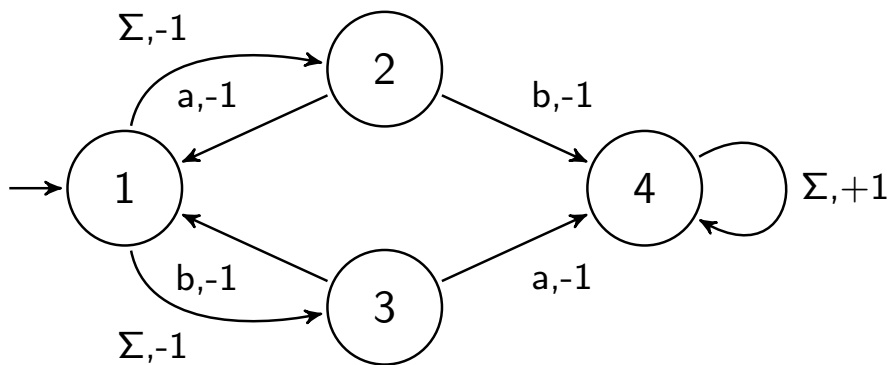
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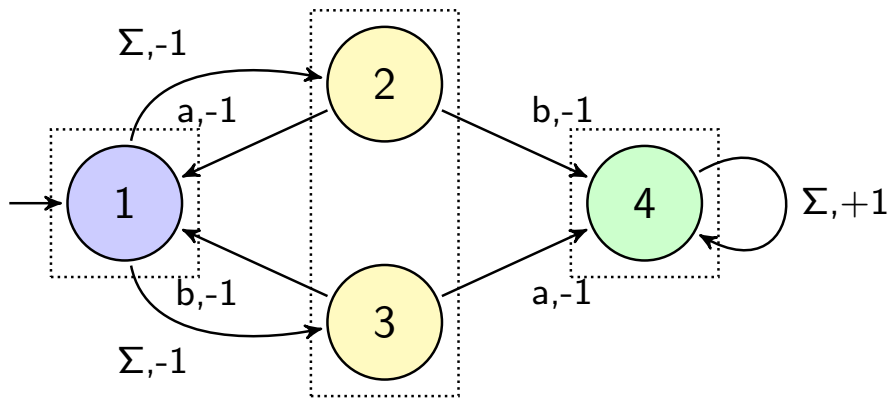
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- we assume Δ is total



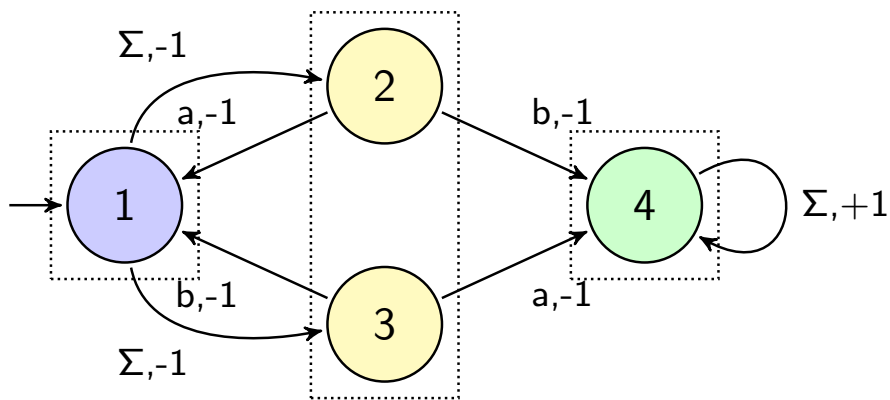
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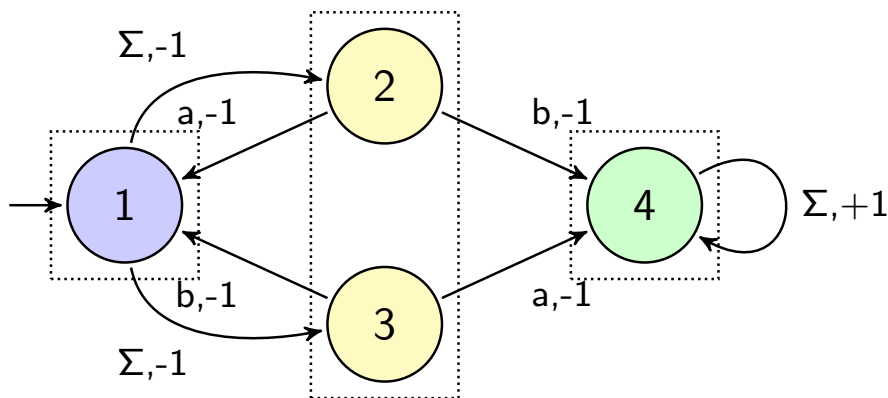
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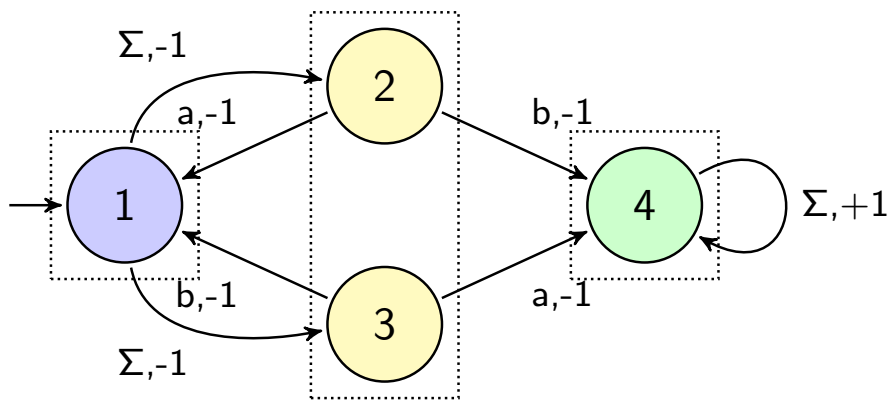
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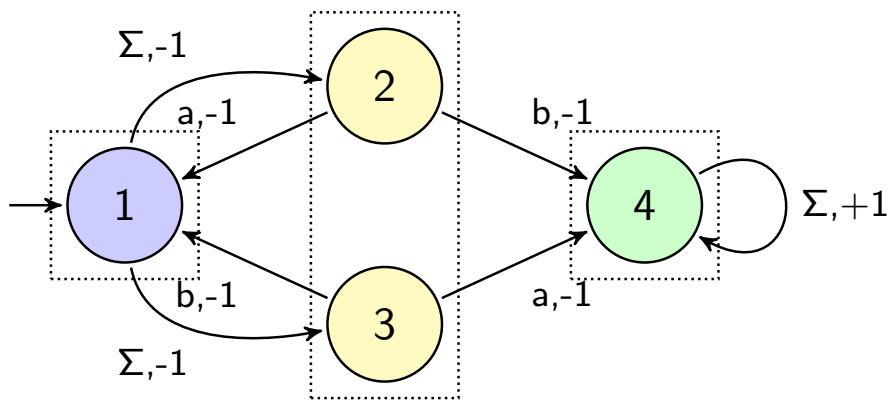
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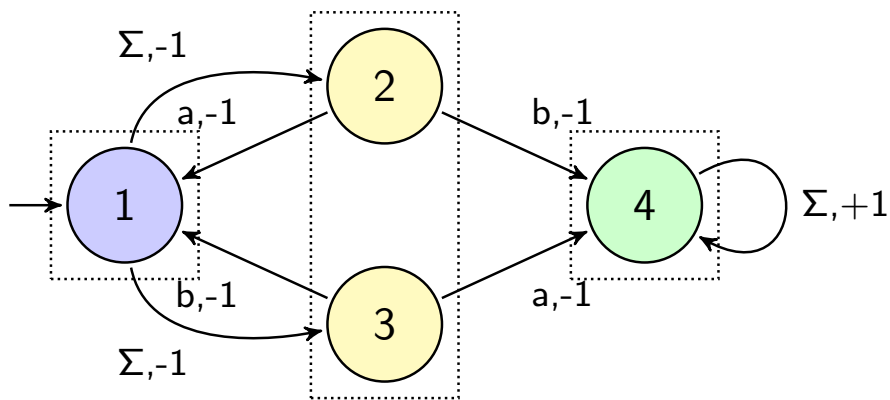
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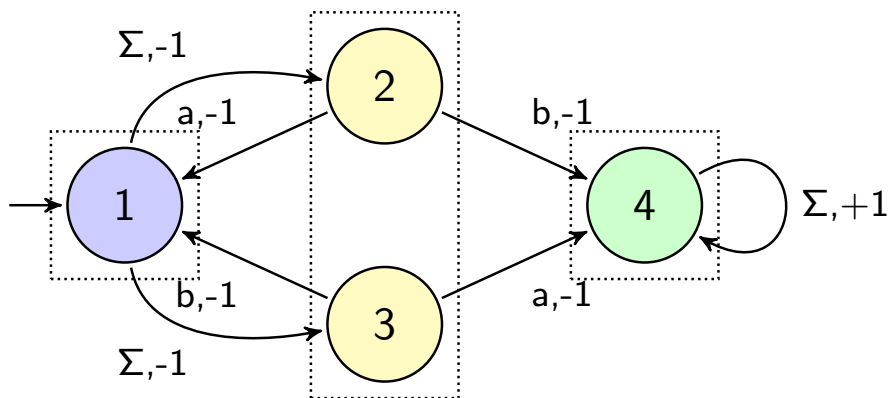
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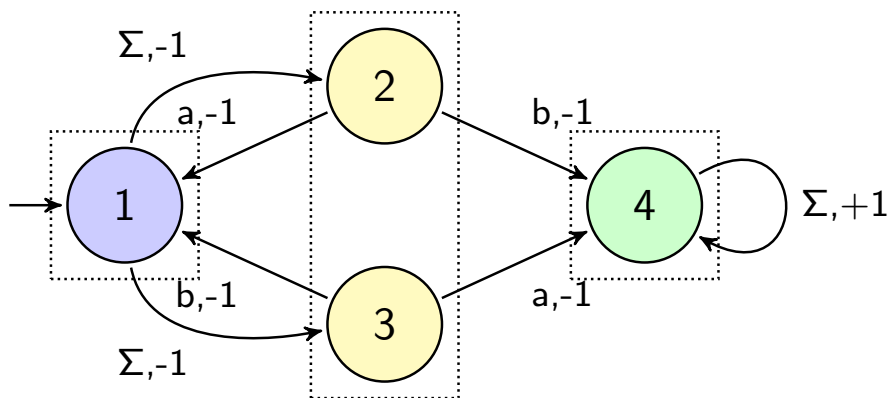
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Some formal definitions

Definition (Strategies for $\exists ve$)

An observation-based strategy for $\exists ve$ is a function from **finite** sequences $(Obs \cdot \Sigma)^* Obs$ to the next action.

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Definition (MP value)

Given a **concrete play** $\pi = q_0 \sigma_0 q_1 \dots \in (Q \cdot \Sigma)^\omega$ and weight function $w : \Delta \rightarrow \mathbb{Z}$, the **MP value** is $\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{i=0}^{n-1} w(q_i, \sigma_i, q_{i+1})$ (\overline{MP} , \underline{MP}).

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Problem (The threshold problem)

Given a threshold $\nu \in \mathbb{N}$, the MPG is won by \exists ve iff she can ensure $MP \geq \nu$ for all concretizations of her witnessed play.^a

^aW.l.o.g. we assume $\nu = 0$.

Motivation

Full-observation (i.e. each state a unique color)

Worst-case analysis of^a

- online metrical task systems,
- finite window online string matching,
- selection with limited storage.

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In the context of reactive synthesis

- MPGs are natural models for systems where we want to optimize the limit-average usage of a resource.
- Partial-observation arises from the fact that most systems have a limited amount of sensors and input data.

What is known?

Theorem (Full-observation [EM79, ZP96])

- *MPGs are **determined**, i.e. if $\exists ve$ doesn't have a winning strategy then $\forall dam$ does (and vice versa).*
- *Positional strategies suffice for either $\forall dam$ or $\exists ve$ to win a MPG.*
- *The threshold problem is in $NP \cap coNP$.*

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Theorem (Partial-observation [DDG⁺10] and [HPR14])

- *$\exists ve$ learns about the state of the game by using (possibly infinite) memory.*
- *The limit does not always exist, i.e. $\overline{MP} \neq \underline{MP}$.*
- *The threshold problem is **undecidable**.*

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Introduced in [CDRR13]

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Key Idea (main construction block)

A **good window** from turn i ...

- Window of fixed size sliding along a play
- \exists ve must see a **local positive MP** before hitting the end of the window

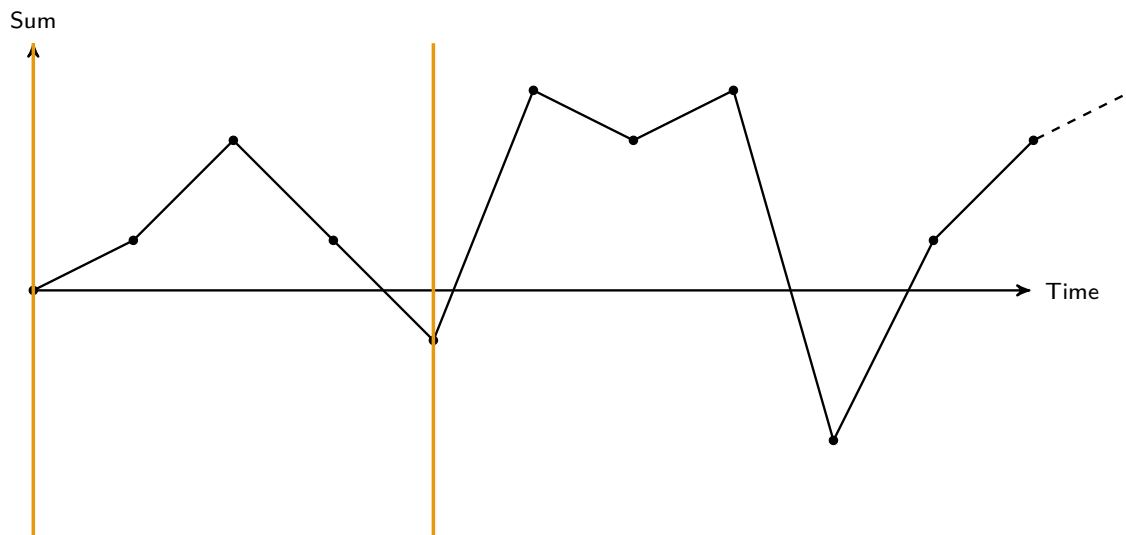
Window MP, threshold zero, maximal window = 4

$$\mathbf{GW}(i, \ell_{\max}) \equiv \exists i \leq \ell_{\max} : w(q_i \dots q_{i+j}) \geq 0$$

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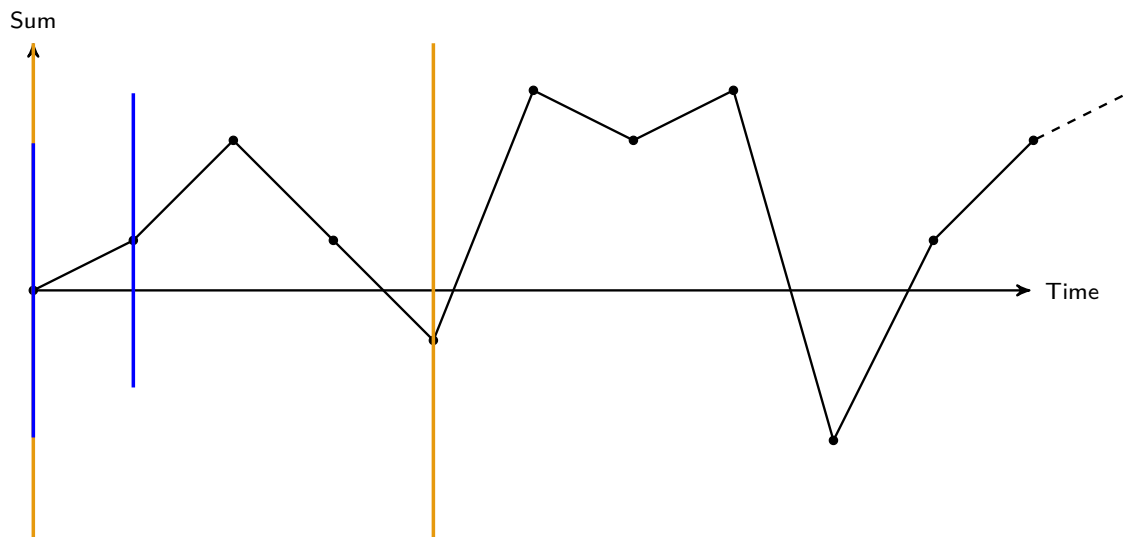
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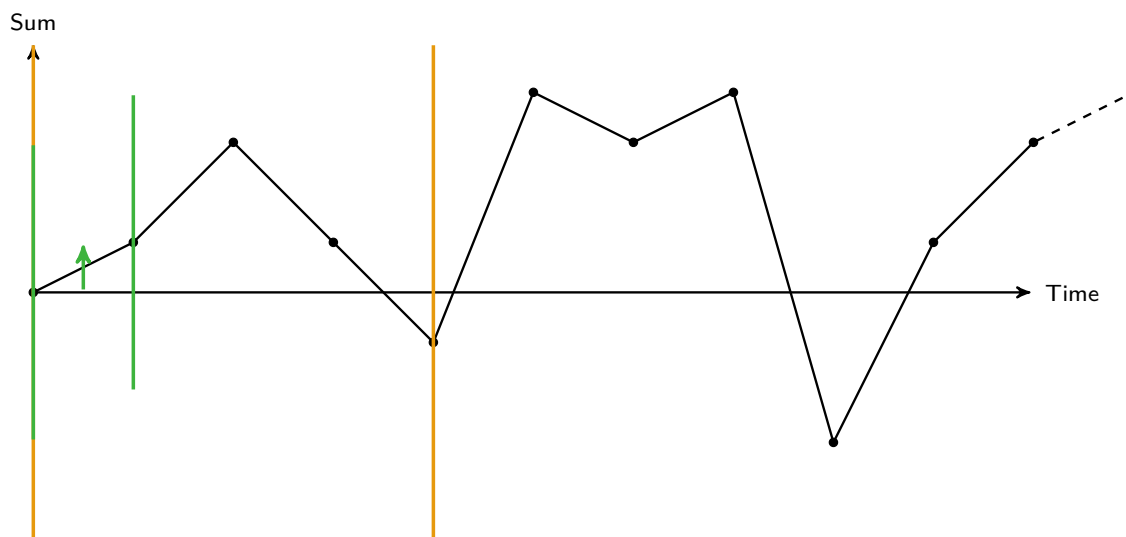
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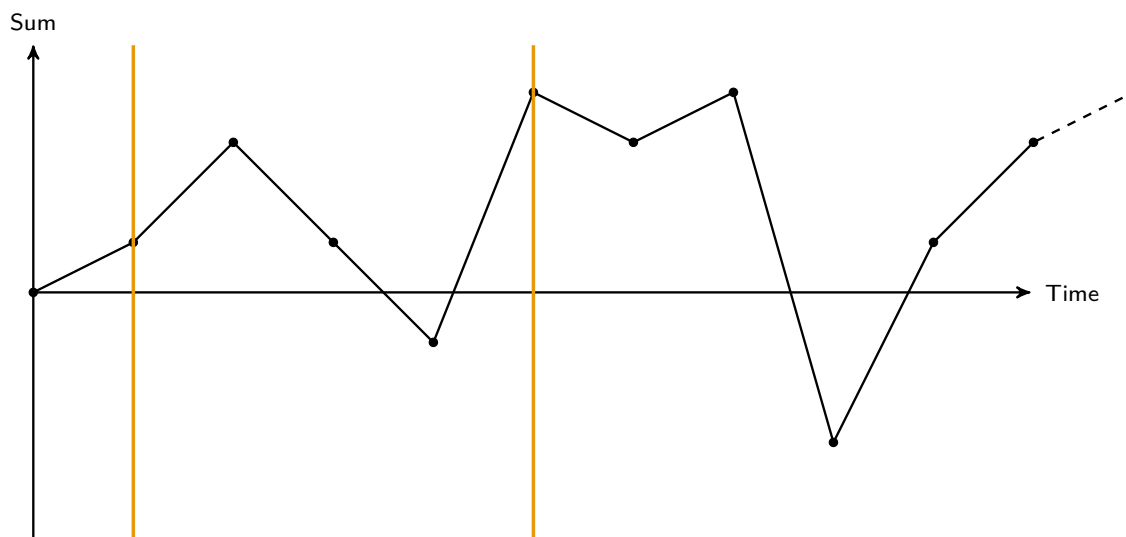
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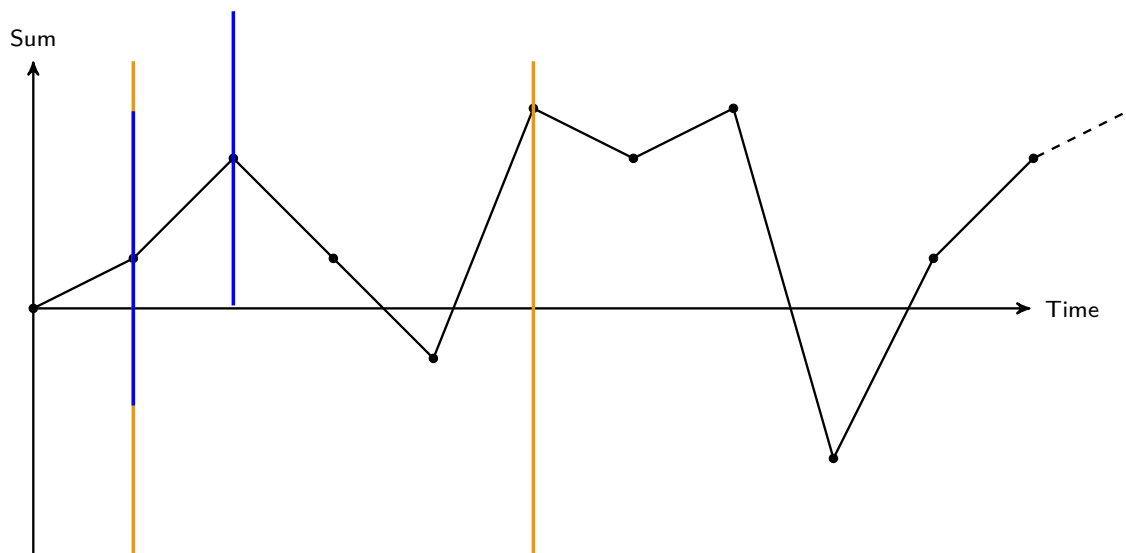
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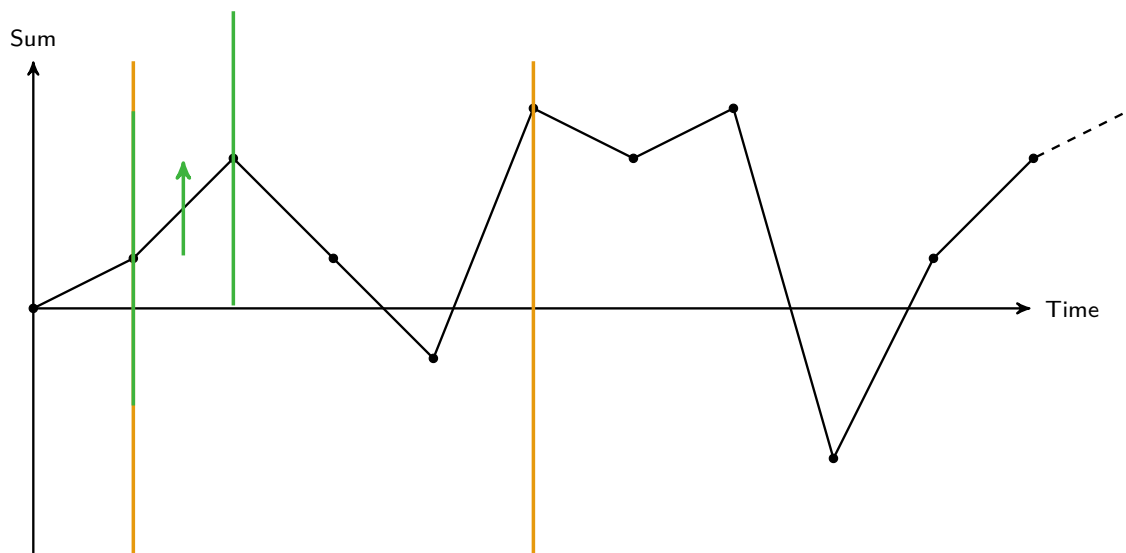
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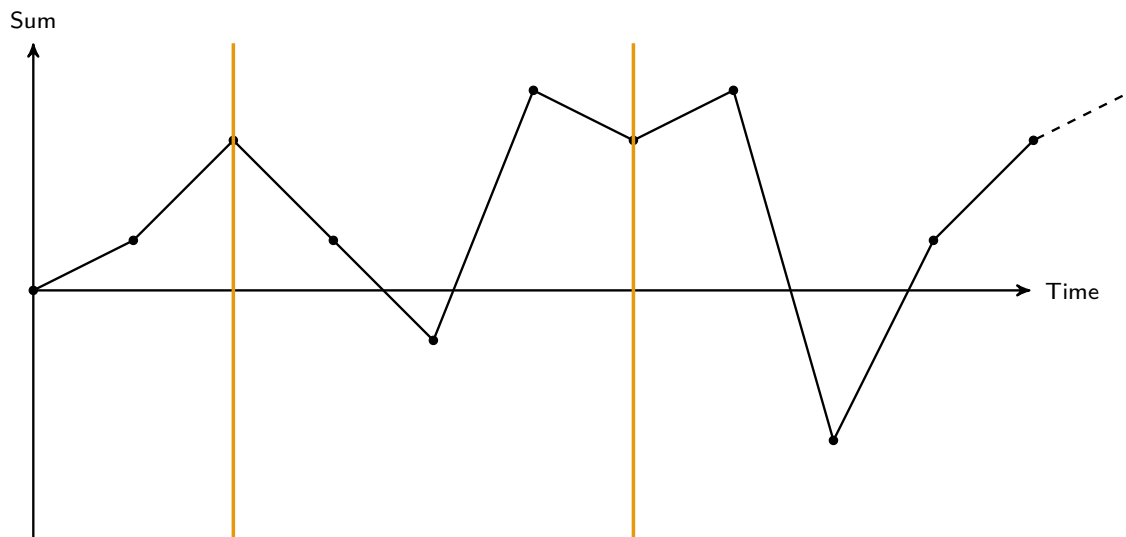
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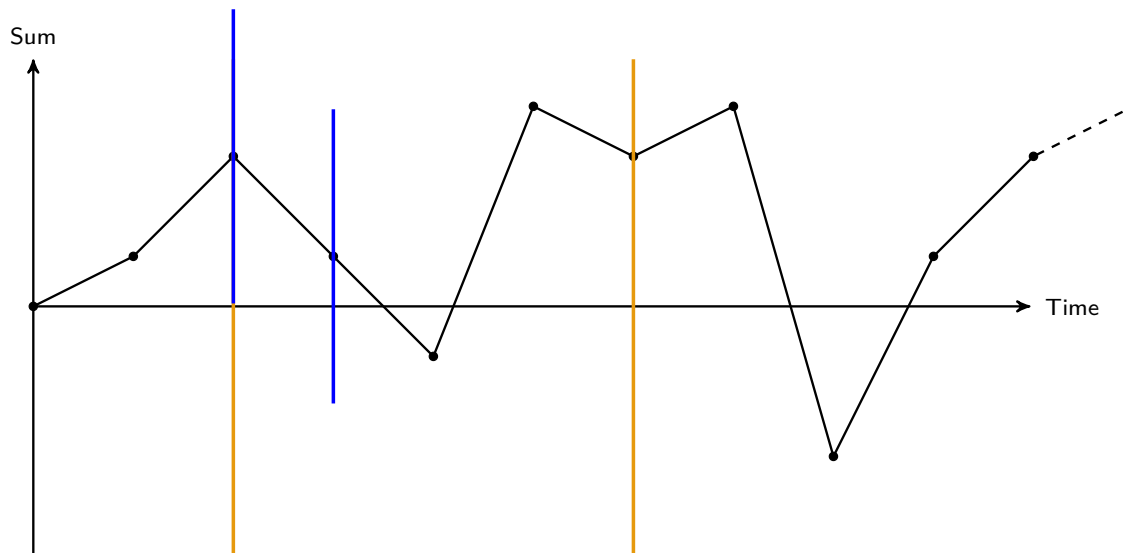
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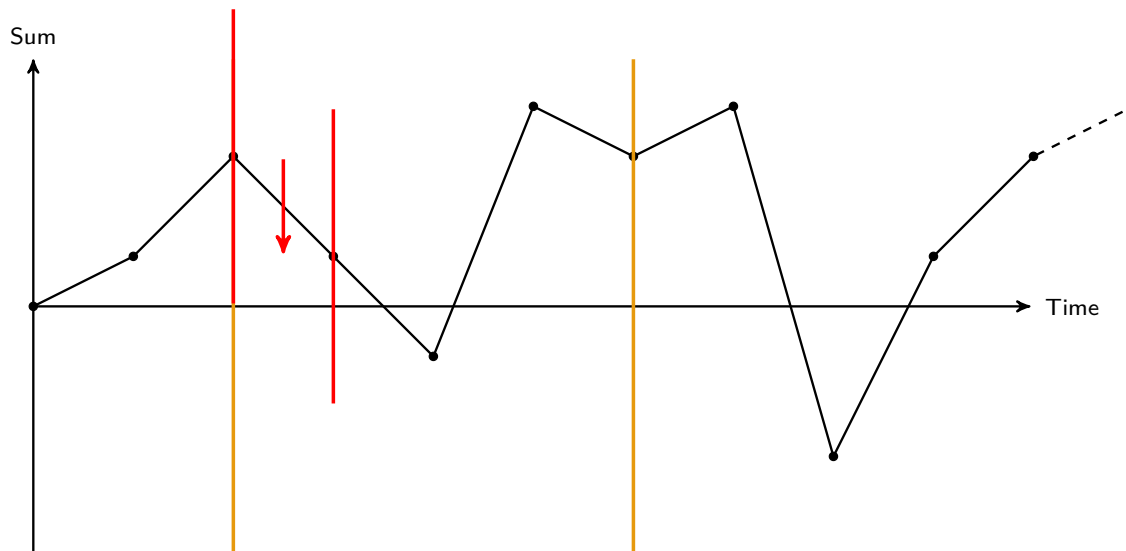
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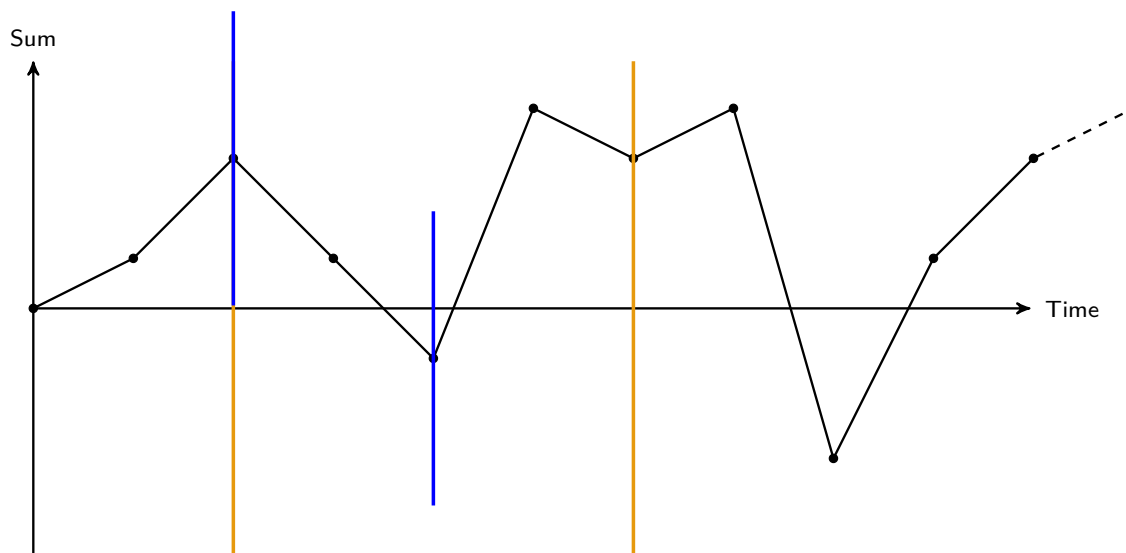
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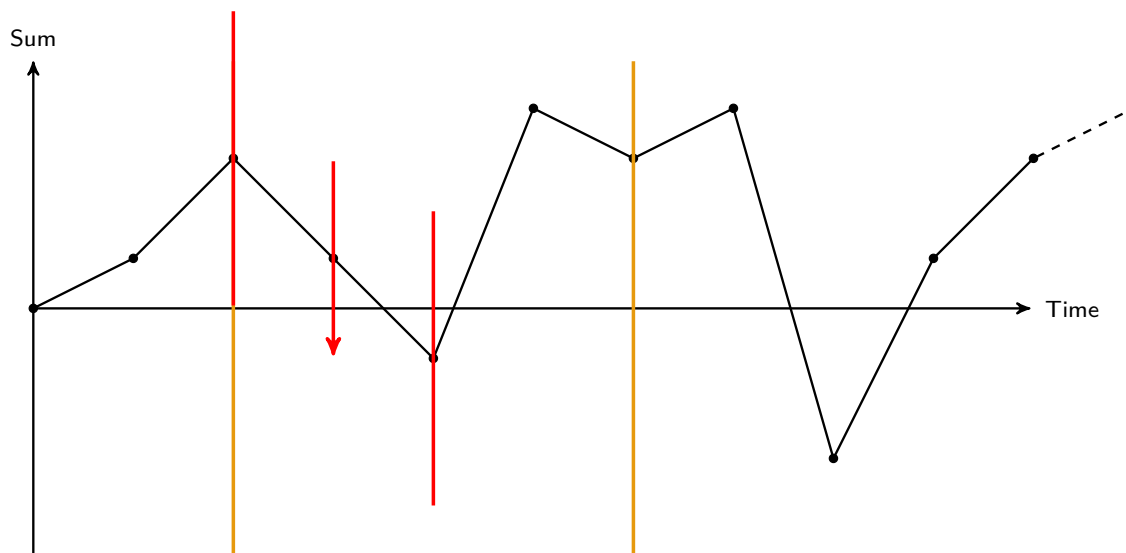
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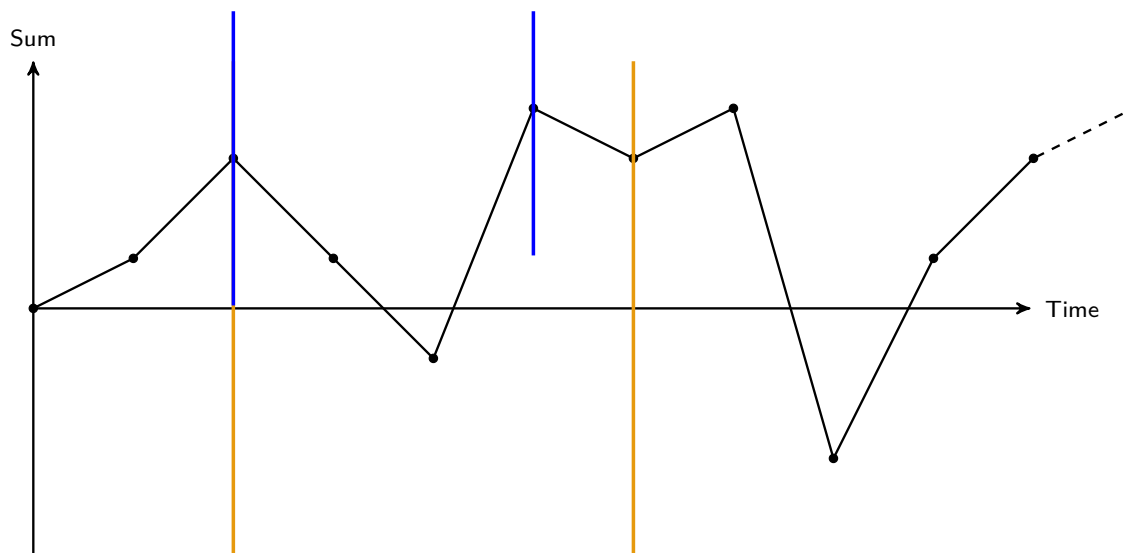
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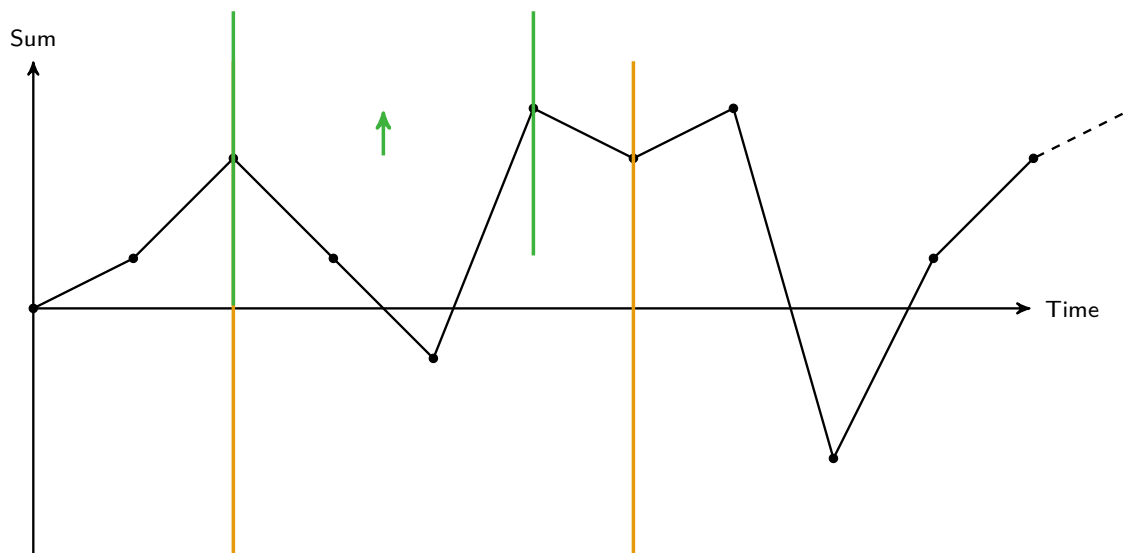
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Multiple variants

Several window objectives were defined in [CDRR13]:

fixed window objectives (FWMP)

using **GW** and a given ℓ_{\max} of polynomial size

- a safety-like WMP objective
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bounded window objectives (BWMP)

quantifying ℓ_{\max} existentially

- safety and co-Büchi WMP objectives

Properties

Theorem (Conservative approximations [CDRR13])

If $\exists ve$ wins any window objective, she wins the MPG for $M \geq 0$. If she wins the MPG for $MP > 0$, $\exists \ell_{\max}$ with which she wins all FWMP.

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Complexity¹

	1-dim	k-dims
FWMP	P-c	$\in \text{EXP}, \text{PSPACE-h}$
BWMP	$\in \text{NP} \cap \text{coNP}, \text{MPG-h}$	NPR-h

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Idea/Motivation

Use the **GW** property of concrete plays to define:

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Main issue

- Where to quantify over concretizations? e.g. for a bounded window objective, should all concretizations have the same ℓ_{\max} ?
- We consider all distinct cases.

PO variants (definitions)

$$\mathbf{GW}(i, \ell_{\max}) \equiv \exists i \leq \ell_{\max} : w(q_i \dots q_{i+j}) \geq 0$$

Using **GW** and a given ℓ_{\max} of polynomial size

Direct Fixed Window (i.e. safety-like)

$$\mathbf{DirFix}(\ell_{\max}) \equiv \forall \pi, \square, \mathbf{GW}(\ell_{\max}) \equiv \square, \forall \pi, \mathbf{GW}(\ell_{\max})$$

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Quantifying ℓ_{\max} existentially: *(Direct) Bounded Window ...*

Categorizing objectives

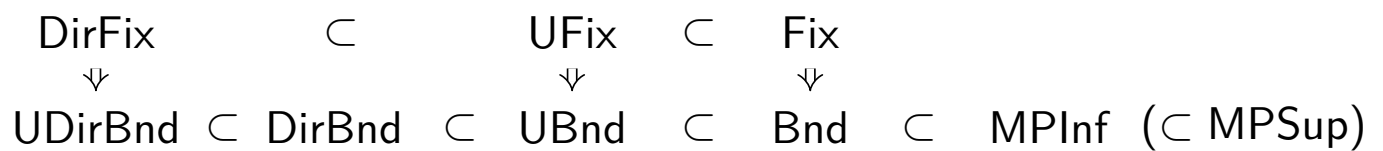


Figure: Implications among the objectives

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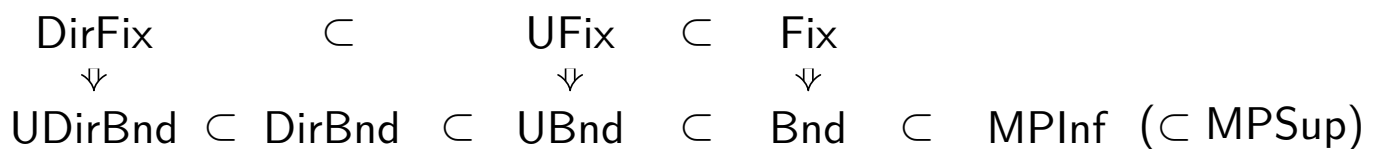


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From weighted automata

Using a reduction from the UNIVERSALITY PROBLEM for weighted automata, we show all bounded window objs. are **undecidable**.

More details on the undecidability...

Weighted automata

A weighted finite automaton $(Q, q_I, \Sigma, \Delta, w)$ realizes a function $\Sigma^* \rightarrow \mathbb{Z}$ by mapping a word x to the minimal (sum) value of its runs on x .

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A weighted automaton is **universal** if $\forall x \in \Sigma^*$ the value assigned by the automaton to x is negative. Undecidability from [ABK11].

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- force $\exists ve$ to play infinitely many $\#$ (a fresh new symbol),
- force her to play $\#$ in intervals of bounded length, and
- when $\exists ve$ plays a word $\#x\#$ with **positive value**, she closes a good window.

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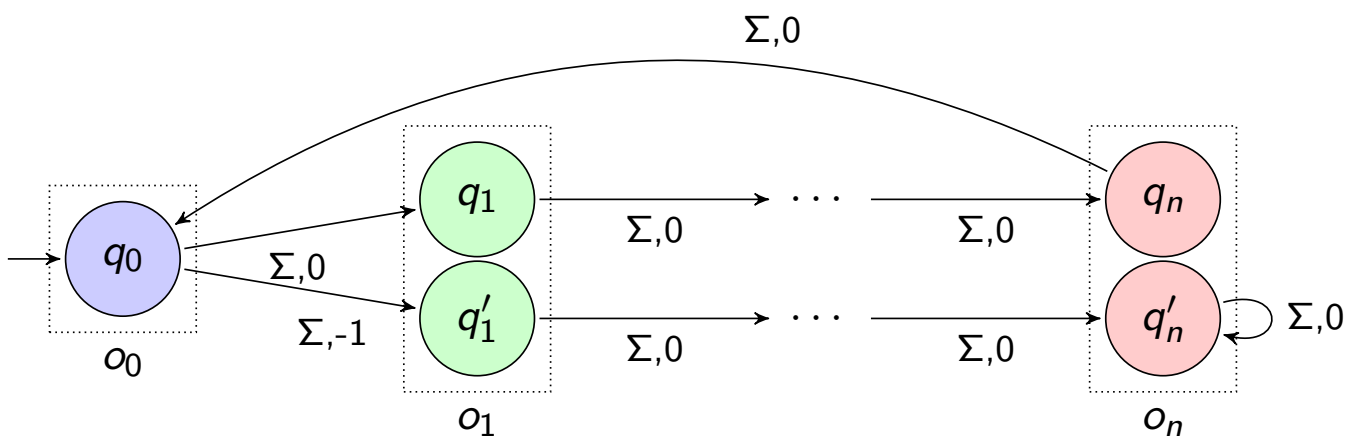
DirFix games: a game on functions

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Direct Fixed Window:

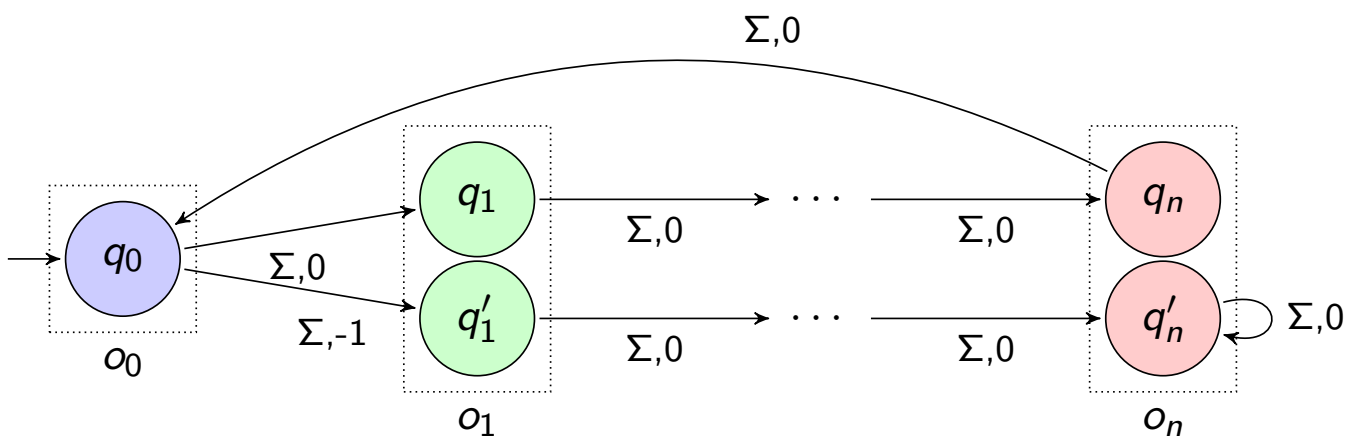
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$${}^1 n = \ell_{\max} + 1$$

DirFix games: $\forall \pi, \square, \mathbf{GW}(\ell_{\max})$



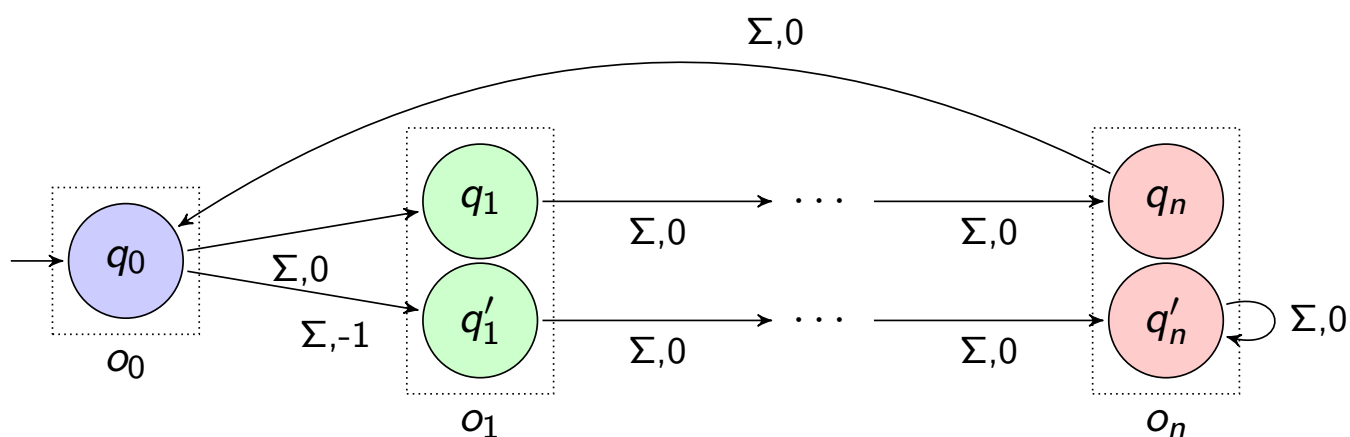
obs:

f. play:

curr. $f \in Q \rightarrow (\{1, \dots, \ell_{\max}\} \rightarrow \{-W \cdot \ell_{\max}, \dots, 0\})$

¹ $n = \ell_{\max} + 1$

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obs: o_0

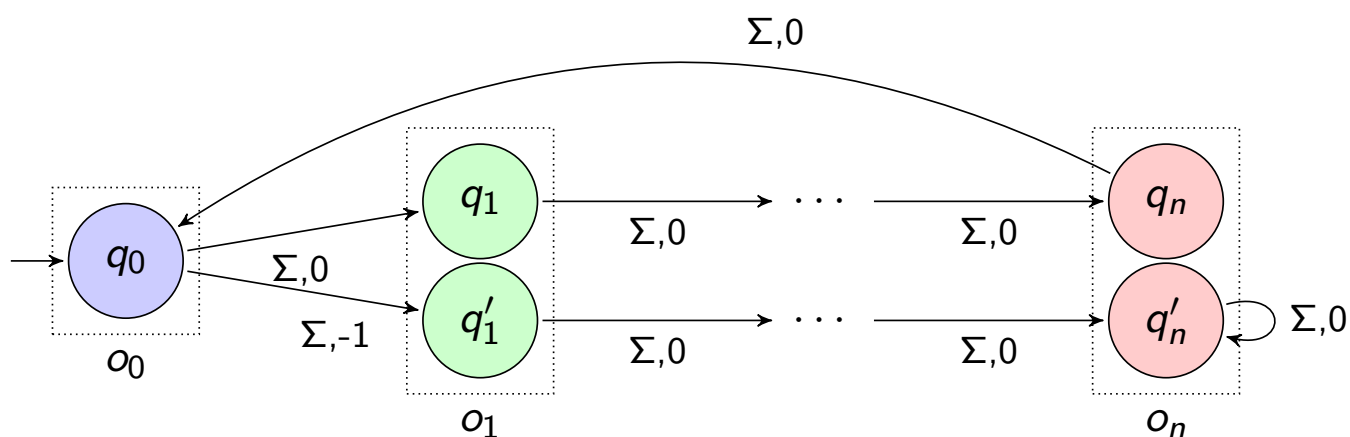
f. play: f_0

curr. $f \in Q \rightarrow (\{1, \dots, \ell_{\max}\} \rightarrow \{-W \cdot \ell_{\max}, \dots, 0\})$

$q_0 \mapsto (i \mapsto 0, \forall i)$

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DirFix games: $\forall \pi, \square, \mathbf{GW}(\ell_{\max})$



obs: $o_0 \sigma_0 o_1$

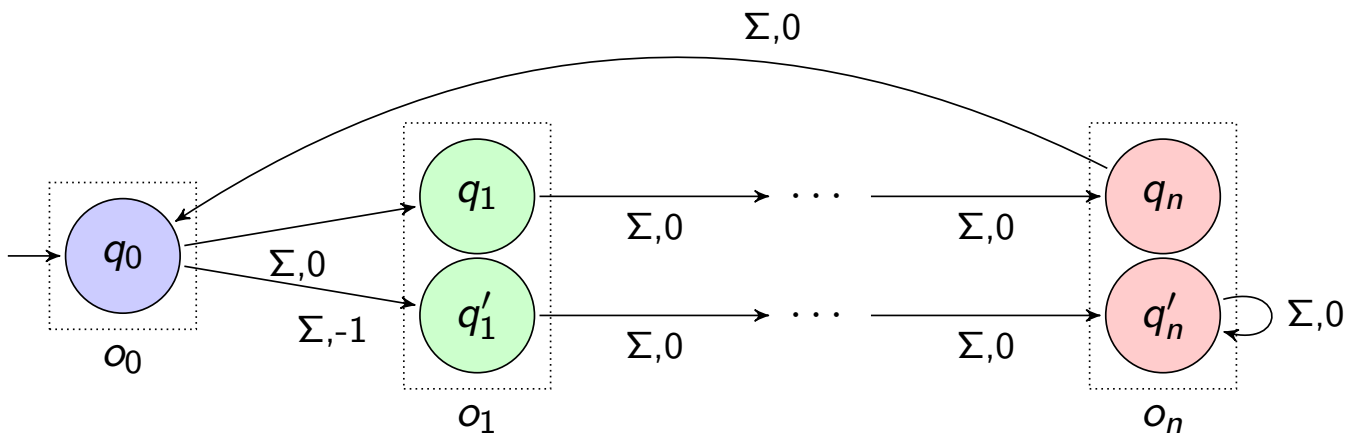
f. play: $f_0 \sigma_0 f_1$

curr. $f \in Q \rightarrow (\{1, \dots, \ell_{\max}\} \rightarrow \{-W \cdot \ell_{\max}, \dots, 0\})$

$q_1 \mapsto (i \mapsto 0, \forall i), q'_1 \mapsto (1 \mapsto -1)$

¹ $n = \ell_{\max} + 1$

DirFix games: $\forall \pi, \square, \mathbf{GW}(\ell_{\max})$



obs: $o_0 \sigma_0 o_1 \sigma_1 o_2$

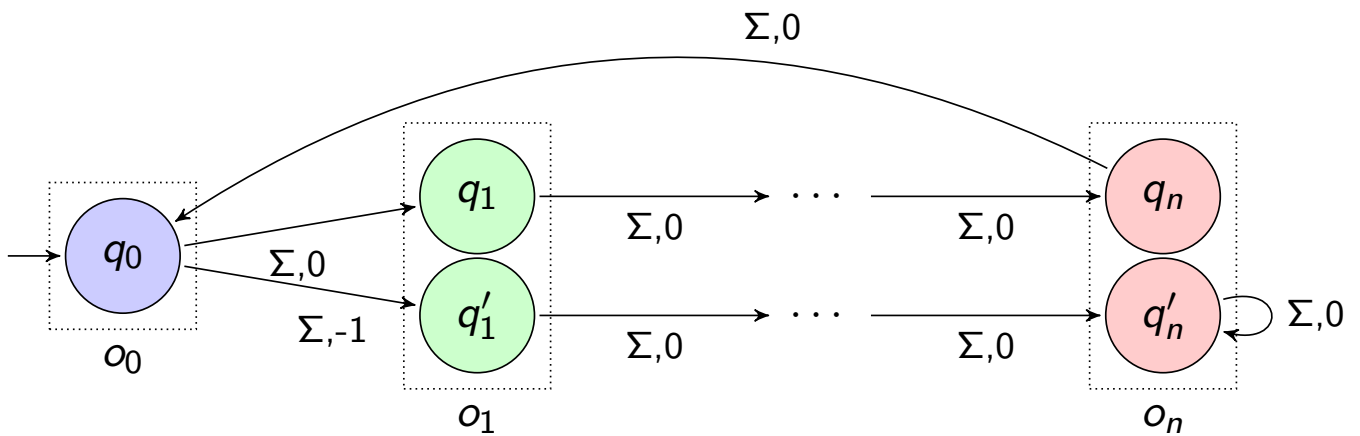
f. play: $f_0 \sigma_0 f_1 \sigma_1 f_2$

curr. $f \in Q \rightarrow (\{1, \dots, \ell_{\max}\} \rightarrow \{-W \cdot \ell_{\max}, \dots, 0\})$

$q_2 \mapsto (i \mapsto 0, \forall i), q'_2 \mapsto (2 \mapsto -1)$

¹ $n = \ell_{\max} + 1$

DirFix games: $\forall \pi, \square, \mathbf{GW}(\ell_{\max})$



obs: $o_0 \sigma_0 o_1 \sigma_1 o_2 \dots \sigma_{n-1} o_n$

f. play: $f_0 \sigma_0 f_1 \sigma_1 f_2 \dots \sigma_{n-1} f_n$

curr. $f \in Q \rightarrow (\{1, \dots, \ell_{\max}\} \rightarrow \{-W \cdot \ell_{\max}, \dots, 0\})$

$q_n \mapsto (i \mapsto 0, \forall i), q'_n \mapsto (n \mapsto -1)$

a window open for $> \ell_{\max}$ steps in q'_n !

¹ $n = \ell_{\max} + 1$

Fix games: using observers

$$\mathbf{GW}(i, \ell_{\max}) \equiv \exists i \leq \ell_{\max} : w(q_i \dots q_{i+j}) \geq 0$$

Fixed Window: $\mathbf{Fix}(\ell_{\max}) \equiv \forall \pi, \diamond, \square, \mathbf{GW}(\ell_{\max})$

Fix games: using observers

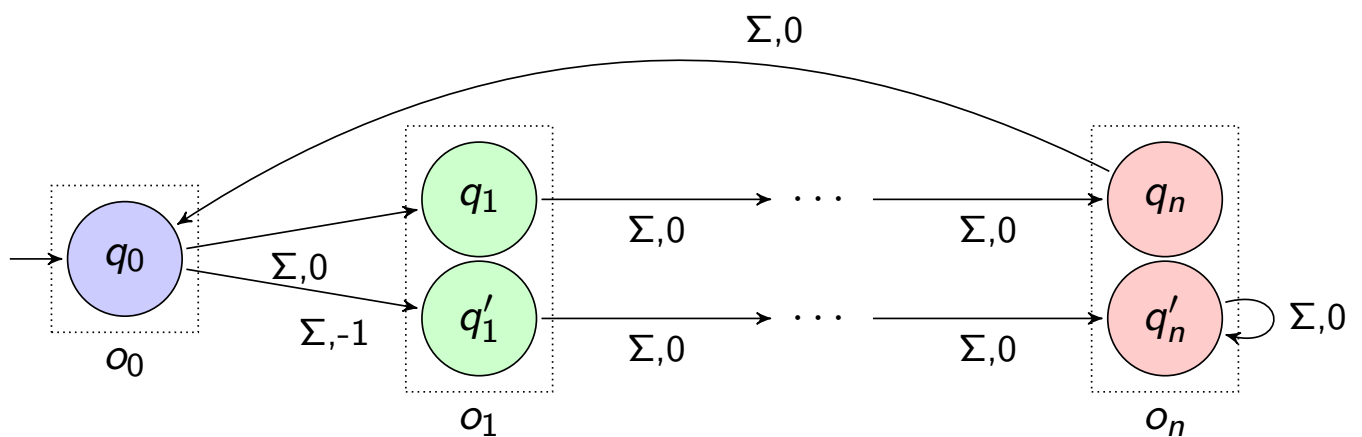
$$\mathbf{GW}(i, \ell_{\max}) \equiv \exists i \leq \ell_{\max} : w(q_i \dots q_{i+j}) \geq 0$$

Fixed Window: $\mathbf{Fix}(\ell_{\max}) \equiv \forall \pi, \diamond, \square, \mathbf{GW}(\ell_{\max})$

The complement of the objective

$\exists \pi, \square, \diamond, \neg \mathbf{GW}(\ell_{\max})$, i.e. there is a concretization with ∞ -ly many bad windows.

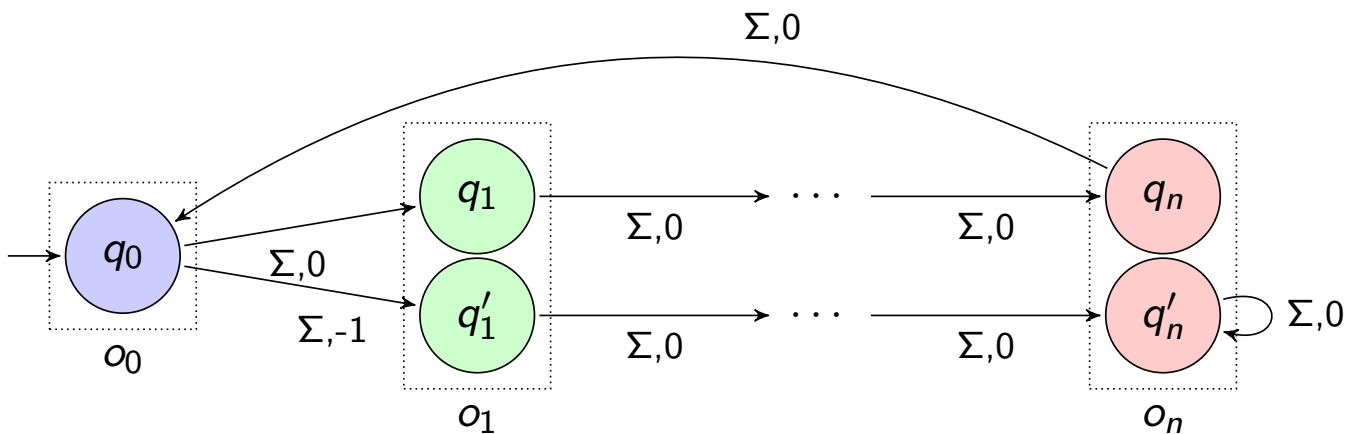
Fix games: $\forall \pi, \diamond, \square, \mathbf{GW}(\ell_{\max})$



Should we play a **co-Büchi** game now?

¹ $n = \ell_{\max} + 1$

Fix games: $\forall\pi, \diamond, \square, \mathbf{GW}(\ell_{\max})$

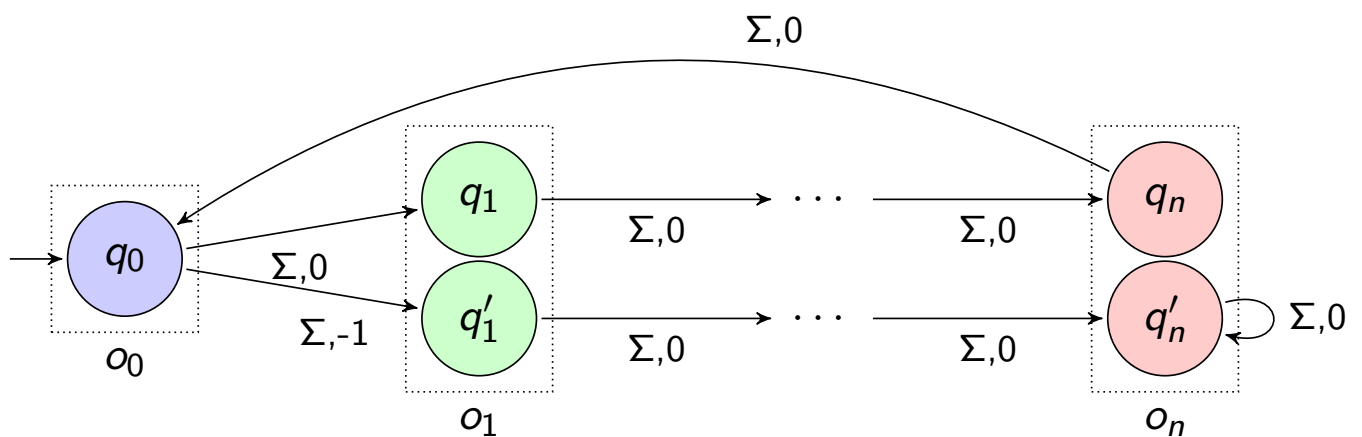


We build a non-det. poly-size Büchi **observer**:

- e.g. q_1, q'_1 are reachable from q_0 reading (Σ, o_1) ;
- states now also encode open window information, i.e. functions $\{1, \dots, \ell_{\max}\} \rightarrow \{-W \cdot \ell_{\max}, \dots, 0\}$;
- states with $f(\ell_{\max}) < 0$ are now accepting.

$$^1 n = \ell_{\max} + 1$$

Fix games: $\forall\pi, \diamond, \square, \mathbf{GW}(\ell_{\max})$



E.g., $q_0, (i \mapsto 0, \forall i)$ non-deterministically transitions to

- $q'_1, (1 \mapsto -1)$ or
- $q_1, (i \mapsto 0, \forall i)$

with input (σ, o_1) .

¹ $n = \ell_{\max} + 1$

UFix games: uniformity

$$\mathbf{GW}(i, \ell_{\max}) \equiv \exists i \leq \ell_{\max} : w(q_i \dots q_{i+j}) \geq 0$$

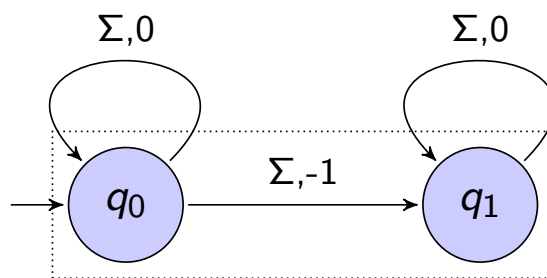
(Uniform) Fixed Window: $\mathbf{UFix}(\ell_{\max}) \equiv \diamond, \forall \pi, \square, \mathbf{GW}(\ell_{\max})$

The complement of the objective

$\square, \exists \pi, \diamond, \neg \mathbf{GW}(\ell_{\max})$, i.e. ∞ -ly many bad windows.

UFix: uniformity (Ex.)

Does \exists ve win for **Fix**? for **UFix**?



UFix: uniformity (merging trick)

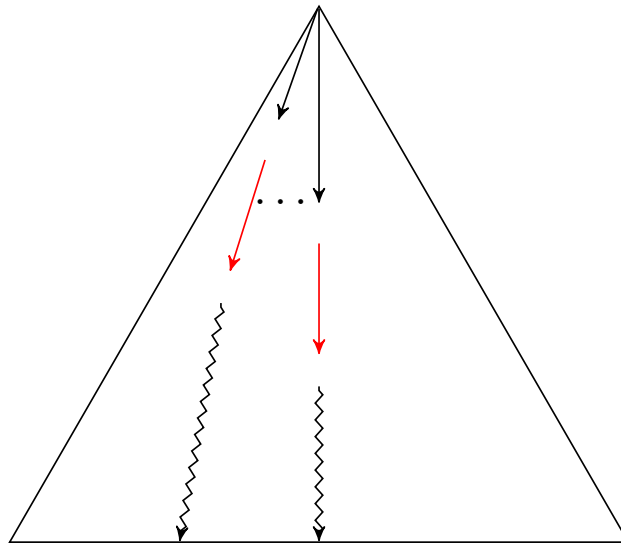
Lemma

*An abstract play is not in **UFix**(ℓ_{max}) iff it has a concretization which merges with infinitely many violating concrete plays.*

UFix: uniformity (merging trick)

Lemma

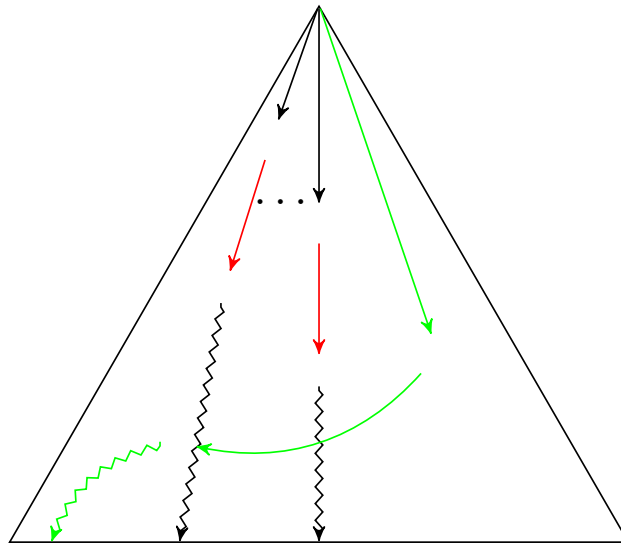
An abstract play is not in $\mathbf{UFix}(\ell_{\max})$ iff it has a concretization which merges with infinitely many violating concrete plays.



UFix: uniformity (merging trick)

Lemma

An abstract play is not in $\mathbf{UFix}(\ell_{\max})$ iff it has a concretization which merges with infinitely many violating concrete plays.



Conclusions





To summarize:

- we extended window objectives to the partial-observation setting,
- obtained a decidable conservative approximation of MPGs (i.e. the FWMP objectives) with **DirFix** games even allowing for an [order on the state-space](#), and
- established tight complexity bounds for their threshold problem.





DirFix	EXP-c
Fix	EXP-c
UFix	EXP-c
Bounded	Undecidable

Still open: full-observation k -dimension BWMP

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