

# Reaching Your Goal Optimally by Playing at Random with no Memory

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# Motivation : game theory for synthesis



## Game theory

Interaction between two  
antagonistic agents :  
environment and controller



## Code synthesis

Correct by  
construction :  
synthesis of  
controller

## Classic approach

Check the correctness  
of a system

# Different sorts of games

## Qualitative games

Reach or avoid some (sequences of) states

## Quantitative games

- ▶ Consider quantitative parameters : energy consumption...
- ▶ Compare distinct strategies

# Different sorts of games

## Qualitative games

Reach or avoid some (sequences of) states

## Quantitative games

- ▶ Consider quantitative parameters : energy consumption...
- ▶ Compare distinct strategies

## Shortest-Path games

- ▶ Combination of a qualitative with a quantitative objective
- ▶ Reach a target with a minimum cost

# Different sorts of games

## Timed games

- ▶ Consider timed issues : receive a message...
- ▶ Infinite games

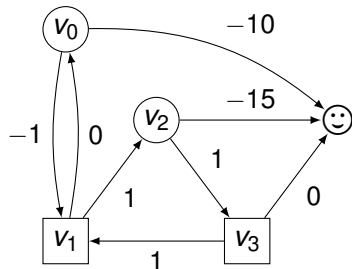
## Weighted Timed games

Combination of timed games and shortest path games.

# Shortest Path Game

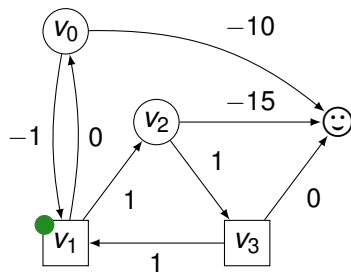
□ Adam    ○ Eve

😊 target (T)



# Shortest Path Game

□ Adam    ○ Eve



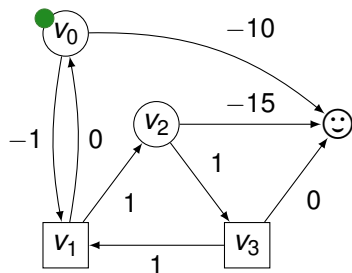
How to play?

Move a token along an edge

$$\pi = v_1$$

# Shortest Path Game

□ Adam    ○ Eve



How to play?

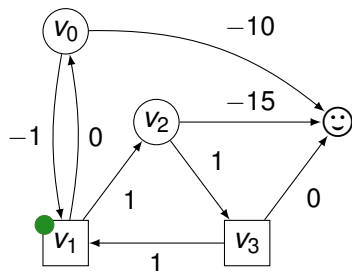
Move a token along an edge

$$\pi = v_1 v_0$$



# Shortest Path Game

□ Adam    ○ Eve



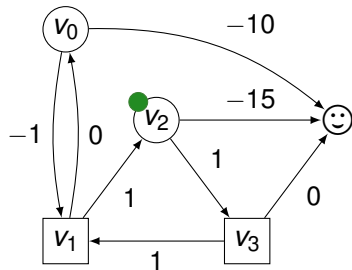
How to play?

Move a token along an edge

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# Shortest Path Game

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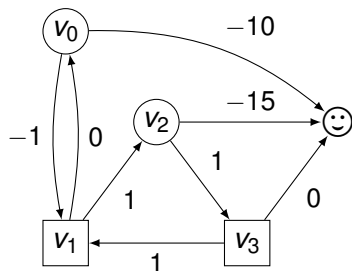
How to play?

Move a token along an edge

$$\pi = v_1 v_0 v_1 v_2$$

# Shortest Path Game

□ Adam    ○ Eve



## Play

Infinite path or reach the target

$$\pi = (v_i)_i \in V^\omega \quad \pi = (v_i)_i \text{☺}$$

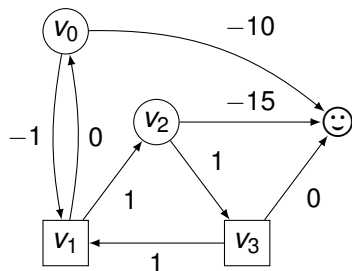
## How to play?

Move a token along an edge

$$\pi = v_1 v_0 v_1 v_2 v_3 \text{☺}$$

# Shortest Path Game

□ Adam    ○ Eve



## Play

Infinite path or reach the target

$$\pi = (v_i)_i \in V^\omega \quad \pi = (v_i)_i \text{😊}$$

## How to play?

Move a token along an edge

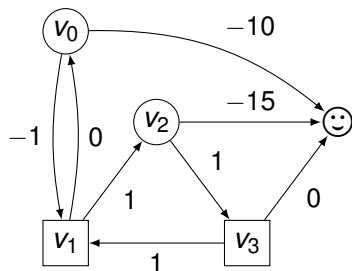
$$\pi = v_1 v_0 v_1 v_2 v_3 \text{😊}$$

## Shortest Path payoff of a play $\pi$

$$\mathbf{SP}(\pi) = \begin{cases} \sum_{i=0}^{n-1} w((\pi_i, \pi_{i+1})) & \text{if } \exists n \text{ (the smallest) s.t. } \pi_n = \text{😊} \\ +\infty & \text{if } \pi \text{ does not reach } \text{😊} \end{cases}$$

# Shortest Path Game

□ Adam    ○ Eve



## Play

Infinite path or reach the target

$$\pi = (v_i)_i \in V^\omega \quad \pi = (v_i)_i \text{😊}$$

## How to play?

Move a token along an edge

$$\pi = v_1 v_0 v_1 v_2 v_3 \text{😊}$$

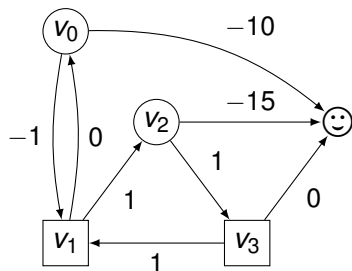
$$\mathbf{SP}(\pi) = 0 + (-1) + 1 + 1 + 0 = 1$$

## Shortest Path payoff of a play $\pi$

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# Shortest Path Game

□ Adam    ○ Eve



## Objectives

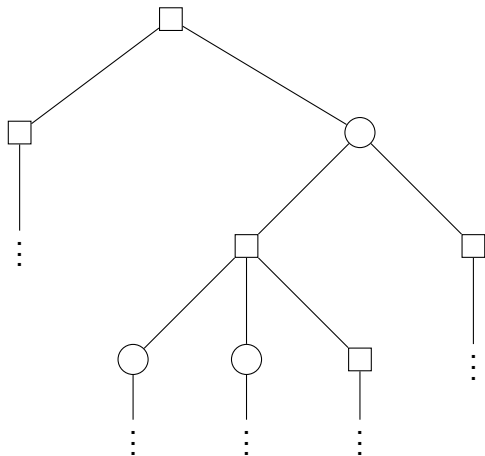
Eve maximise the payoff

Adam minimise the payoff

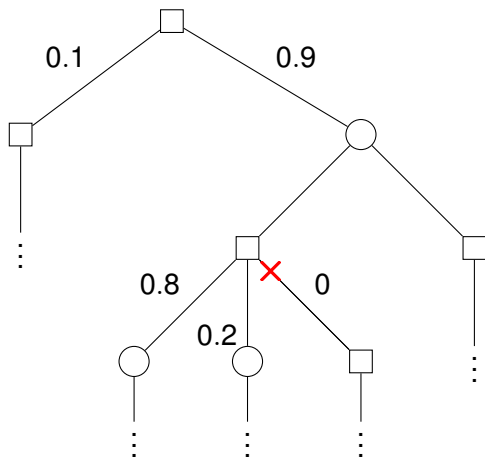
## Shortest Path payoff of a play $\pi$

$$\mathbf{SP}(\pi) = \begin{cases} \sum_{i=0}^{n-1} w((\pi_i, \pi_{i+1})) & \text{if } \exists n \text{ (the smallest) s.t. } \pi_n = \text{smiley face} \\ +\infty & \text{if } \pi \text{ does not reach smiley face} \end{cases}$$

# Strategies for Adam



# Strategies for Adam



A strategy

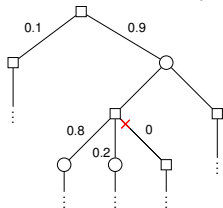
$$\sigma : V^* V_{Adam} \rightarrow \Delta(V)$$



# Strategies for Adam

## Infinite memory

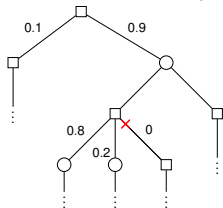
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# Strategies for Adam

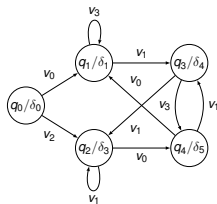
## Infinite memory

$$\sigma : V^* V_{Adam} \rightarrow \Delta(V)$$



## Finite memory

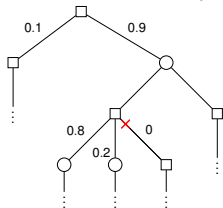
### Moore machine



# Strategies for Adam

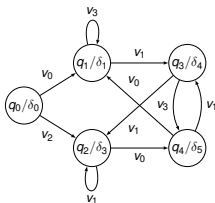
## Infinite memory

$$\sigma : V^* V_{Adam} \rightarrow \Delta(V)$$



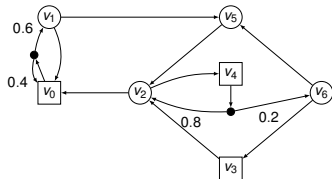
## Finite memory

### Moore machine



## Memoryless

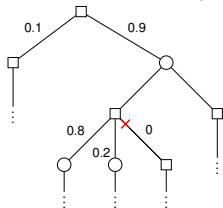
$$\sigma : V_{Adam} \rightarrow \Delta(V)$$



# Strategies for Adam

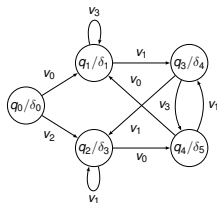
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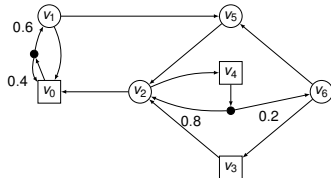
## Finite memory

### Moore machine



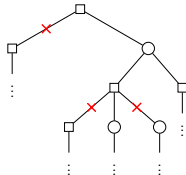
## Memoryless

$$\sigma : V_{Adam} \rightarrow \Delta(V)$$



## Deterministic

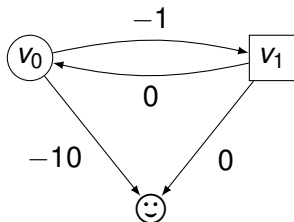
$$\sigma : V^* V_{Adam} \rightarrow V$$



# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$

$\sigma$  Adam     $\tau$  Eve

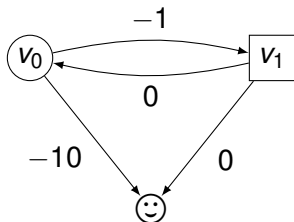


Value

$$\overline{\text{dVal}}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\mathbf{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$

# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$



$\sigma$  Adam     $\tau$  Eve

## Value

$$\overline{\text{dVal}}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\mathbf{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$

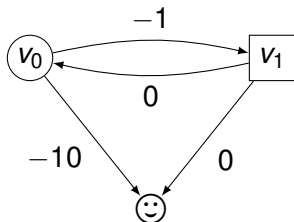
## Determinacy

$$\text{dVal}(v) = \overline{\text{dVal}}(v) = \underline{\text{dVal}}(v)$$

# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$

$\sigma$  Adam     $\tau$  Eve



Value

$$dVal(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\mathbf{SP}(\text{Play}(v, \sigma, \tau))}_{dVal^{\sigma}(v)}$$

## Bellman equation

$$dVal(v) = \begin{cases} 0 & \text{if } v = \text{☺} \\ \max_{v'} (w(v, v') + dVal(v')) & \text{if } v \in V_{\text{Eve}} \\ \min_{v'} (w(v, v') + dVal(v')) & \text{if } v \in V_{\text{Adam}} \end{cases}$$

*Pseudopolynomial iterative algorithm to solve total-payoff games and min-cost reachability games*, T. Brihaye, G. Geeraerts, A. Haddad and B. Monmege, 2016, Acta Informatica

# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$

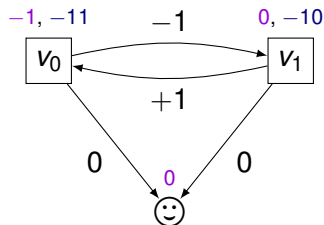
$\sigma$  Adam     $\tau$  Eve

## Value

$$d\text{Val}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\mathbf{SP}(\text{Play}(v, \sigma, \tau))}_{d\text{Val}^{\sigma}(v)}$$

## Unicity

Bellman equation may have many solutions.



## Bellman equation

$$d\text{Val}(v) = \begin{cases} 0 & \text{if } v = \text{smiley face} \\ \max_{v'} (w(v, v') + d\text{Val}(v')) & \text{if } v \in V_{\text{Eve}} \\ \min_{v'} (w(v, v') + d\text{Val}(v')) & \text{if } v \in V_{\text{Adam}} \end{cases}$$

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# Deterministic Strategies

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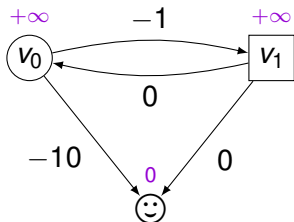
$\sigma$  Adam     $\tau$  Eve

## Value

$$\text{dVal}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\text{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$

## Value iteration

- Compute dVal as a greatest fixpoint



## Bellman equation

$$\text{dVal}(v) = \begin{cases} 0 & \text{if } v = \text{smiley} \\ \max_{v'} (w(v, v') + \text{dVal}(v')) & \text{if } v \in V_{\text{Eve}} \\ \min_{v'} (w(v, v') + \text{dVal}(v')) & \text{if } v \in V_{\text{Adam}} \end{cases}$$

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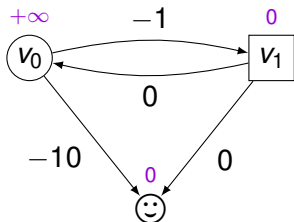
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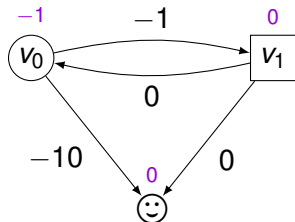
Bellman equation

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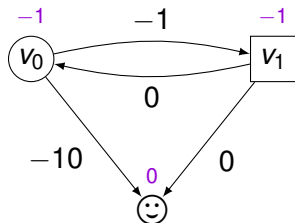
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## Value iteration

- Compute dVal as a greatest fixpoint

## Bellman equation

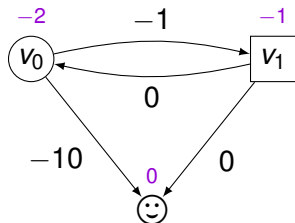
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Value

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Value iteration

- Compute dVal as a greatest fixpoint

Bellman equation

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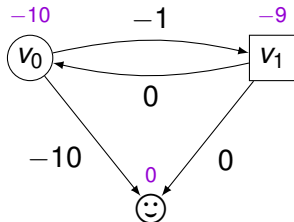
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Value iteration

- Compute dVal as a greatest fixpoint



Bellman equation

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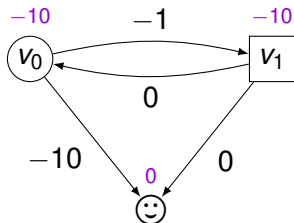
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Value iteration

- Compute dVal as a greatest fixpoint



Bellman equation

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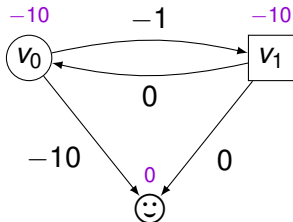
# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$

$\sigma$  Adam     $\tau$  Eve

## Value

$$\text{dVal}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\text{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$



## Value iteration

- ▶ Compute dVal as a greatest fixpoint
- ▶ Complexity: pseudo-polynomial

## Bellman equation

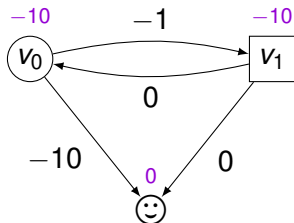
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# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$



$\sigma$  Adam     $\tau$  Eve

Value

$$\text{dVal}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\mathbf{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$

Optimal strategy

$$\text{dVal}^{\sigma^*}(v) \leq \text{dVal}(v)$$

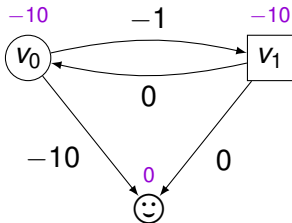
# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$

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Value

$$\text{dVal}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\mathbf{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$



Optimal strategy for Adam

An optimal strategy for Adam may require finite memory.

Optimal strategy

$$\text{dVal}^{\sigma^*}(v) \leq \text{dVal}(v)$$

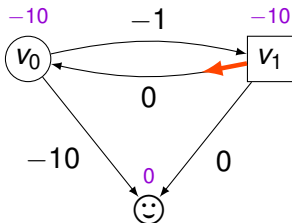
# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$

$\sigma$  Adam     $\tau$  Eve

Value

$$\text{dVal}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\mathbf{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$



Optimal strategy for Adam

Switching strategy:

- ▶  $\sigma_1$  : reach negative cycle

Optimal strategy

$$\text{dVal}^{\sigma^*}(v) \leq \text{dVal}(v)$$

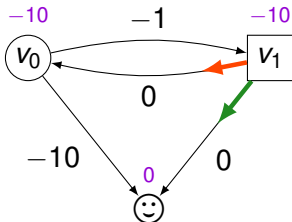
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$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$

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Value

$$\text{dVal}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\text{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$



Optimal strategy for Adam

Switching strategy:

- ▶  $\sigma_1$  : reach negative cycle
- ▶  $\sigma_2$  : reach 😊

Optimal strategy

$$\text{dVal}^{\sigma^*}(v) \leq \text{dVal}(v)$$

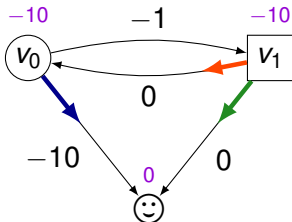
# Deterministic Strategies

$$\sigma : V^* \times V_{\text{Adam}} \rightarrow V$$

$\sigma$  Adam     $\tau$  Eve

Value

$$\text{dVal}(v) = \inf_{\sigma} \sup_{\tau} \underbrace{\text{SP}(\text{Play}(v, \sigma, \tau))}_{\text{dVal}^{\sigma}(v)}$$



Optimal strategy for Adam

Switching strategy:

- ▶  $\sigma_1$  : reach negative cycle
- ▶  $\sigma_2$  : reach 😊

Optimal strategy

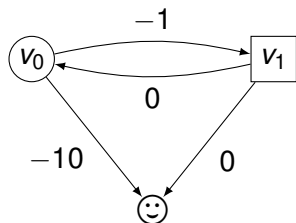
$$\text{dVal}^{\sigma^*}(v) \leq \text{dVal}(v)$$

Optimal strategy for Eve

Eve has a memoryless optimal strategy.

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



$\sigma$  Adam     $\tau$  Eve

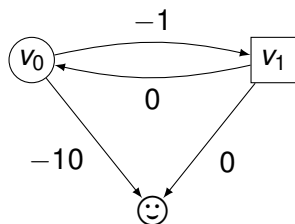
Value

$$\overline{\text{mVal}}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



$\square \sigma$  Adam     $\circ \tau$  Eve

Value

$$\overline{\text{mVal}}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

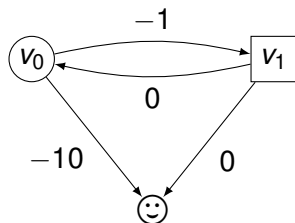
$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$

Determinacy

$$\text{mVal}(v) = \overline{\text{mVal}}(v) = \underline{\text{mVal}}(v)$$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



$\sigma$  Adam     $\tau$  Eve

Value

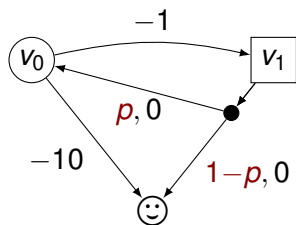
$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$



# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



$\sigma$  Adam     $\tau$  Eve

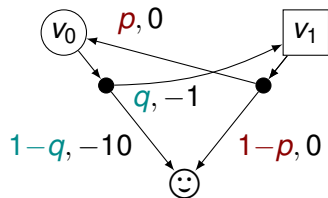
Value

$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



$\sigma$  Adam     $\tau$  Eve

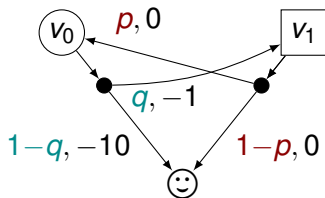
Value

$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



$\sigma$  Adam     $\tau$  Eve

Value

$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

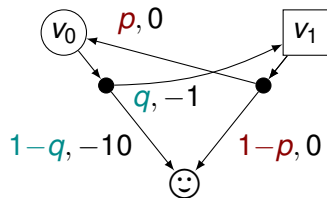
$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$

## Bellman equation in a Markov Chain

$$\text{mVal}^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v')(w(v, v') + \text{mVal}^{\sigma, \tau}(v'))$$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



$\sigma$  Adam     $\tau$  Eve

Value

$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$

Unicity

A unique fixpoint :  $\text{mVal}^{\sigma, \tau}$

## Bellman equation in a Markov Chain

$$\text{mVal}^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v')(w(v, v') + \text{mVal}^{\sigma, \tau}(v'))$$

# Memoryless strategies

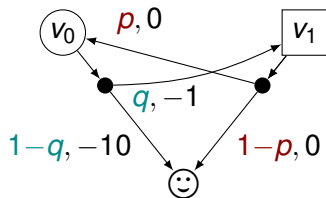
$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$

$\sigma$  Adam     $\tau$  Eve

Value

$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$



Compute  $\text{mVal}^{\sigma, \tau}$

$$\text{mVal}^{\sigma, \tau}(v_1) = p \times \text{mVal}^{\sigma, \tau}(v_0)$$

$$\text{mVal}^{\sigma, \tau}(v_0) = q(\text{mVal}^{\sigma, \tau}(v_1) - 1) - 10(1 - q)$$

Bellman equation in a Markov Chain

$$\text{mVal}^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v')(w(v, v') + \text{mVal}^{\sigma, \tau}(v'))$$

# Memoryless strategies

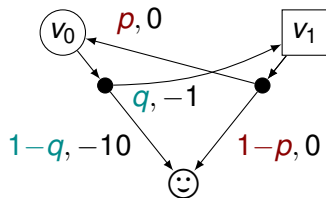
$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$

$\sigma$  Adam     $\tau$  Eve

Value

$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$



Compute  $\text{mVal}^{\sigma, \tau}$

$$\text{mVal}^{\sigma, \tau}(v_1) = p \frac{-q-10(1-q)}{1-pq}$$

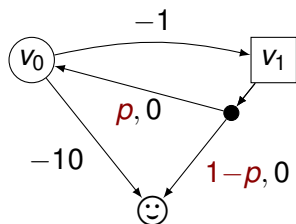
$$\text{mVal}^{\sigma, \tau}(v_0) = \frac{-q-10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$\text{mVal}^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v')(w(v, v') + \text{mVal}^{\sigma, \tau}(v'))$$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



Compute  $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v') (w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{mVal^{\sigma}(v)}$

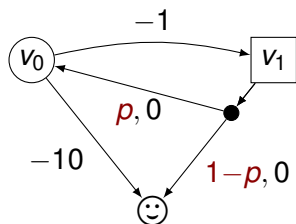
Compute  $mVal^{\sigma}$

► If  $p < \frac{9}{10}$ , then  $q = 1$  :

► If  $p \geq \frac{9}{10}$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



Compute  $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v')(w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{mVal^{\sigma}(v)}$

Compute  $mVal^{\sigma}$

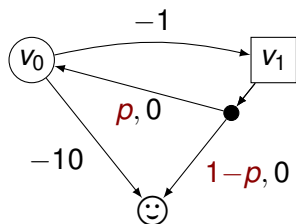
► If  $p < \frac{9}{10}$ , then  $q = 1$ :

► If  $p \geq \frac{9}{10}$ , then  $q = 0$ :



# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



## Compute $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

## Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v') (w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$mVal^{\sigma}(v)$

## Compute $mVal^{\sigma}$

- ▶ If  $p < \frac{9}{10}$ , then  $q = 1$ :

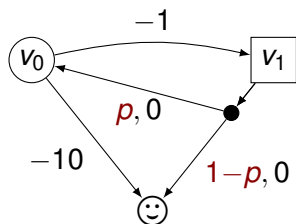
$$mVal^{\sigma}(v_1) = \frac{-p}{1-p}$$

$$mVal^{\sigma}(v_0) = \frac{-1}{1-p}$$

- ▶ If  $p \geq \frac{9}{10}$ , then  $q = 0$ :

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



Compute  $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v') (w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{mVal^{\sigma}(v)}$

Compute  $mVal^{\sigma}$

- ▶ If  $p < \frac{9}{10}$ , then  $q = 1$ :

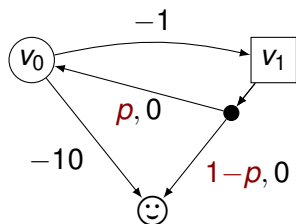
$$mVal^{\sigma}(v_1) = \frac{-p}{1-p}$$

$$mVal^{\sigma}(v_0) = \frac{-1}{1-p}$$

- ▶ If  $p \geq \frac{9}{10}$ , then  $q = 0$ :

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



Compute  $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v') (w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$mVal^{\sigma}(v)$

Compute  $mVal^{\sigma}$

- ▶ If  $p < \frac{9}{10}$ , then  $q = 1$ :

$$mVal^{\sigma}(v_1) = \frac{-p}{1-p}$$

$$mVal^{\sigma}(v_0) = \frac{-1}{1-p}$$

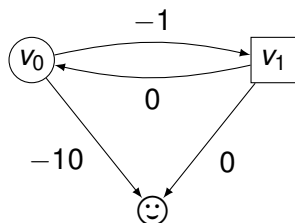
- ▶ If  $p \geq \frac{9}{10}$ , then  $q = 0$ :

$$mVal^{\sigma}(v_1) = -10p$$

$$mVal^{\sigma}(v_0) = -10$$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



Compute  $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v') (w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$mVal^{\sigma}(v)$

Compute  $mVal^{\sigma}$

- ▶ If  $p < \frac{9}{10}$ , then  $q = 1$ :

$$mVal^{\sigma}(v_1) = \frac{-p}{1-p}$$

$$mVal^{\sigma}(v_0) = \frac{-1}{1-p}$$

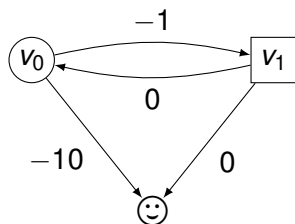
- ▶ If  $p \geq \frac{9}{10}$ , then  $q = 0$ :

$$mVal^{\sigma}(v_1) = -10p$$

$$mVal^{\sigma}(v_0) = -10$$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



Compute  $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v') (w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

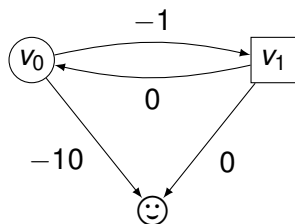
$mVal^{\sigma}(v)$

Compute  $mVal^{\sigma}$

- ▶ If  $p < \frac{9}{10}$ , then  $q = 1$ :  
 $mVal^{\sigma}(v_1) = \frac{-p}{1-p} > -9$   
 $mVal^{\sigma}(v_0) = \frac{-1}{1-p}$
- ▶ If  $p \geq \frac{9}{10}$ , then  $q = 0$ :  
 $mVal^{\sigma}(v_1) = -10p$   
 $mVal^{\sigma}(v_0) = -10$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



Compute  $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v') (w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{mVal^{\sigma}(v)}$

Compute  $mVal^{\sigma}$

- ▶ If  $p < \frac{9}{10}$ , then  $q = 1$ :

$$mVal^{\sigma}(v_1) = \frac{-p}{1-p} > -9$$

$$mVal^{\sigma}(v_0) = \frac{-1}{1-p} > -10$$

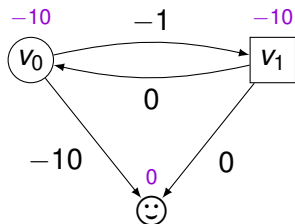
- ▶ If  $p \geq \frac{9}{10}$ , then  $q = 0$ :

$$mVal^{\sigma}(v_1) = -10p$$

$$mVal^{\sigma}(v_0) = -10$$

# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$



Compute  $mVal^{\sigma, \tau}$

$$mVal^{\sigma, \tau}(v_1) = p \frac{-q - 10(1-q)}{1-pq}$$

$$mVal^{\sigma, \tau}(v_0) = \frac{-q - 10(1-q)}{1-pq}$$

Bellman equation in a Markov Chain

$$mVal^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v') (w(v, v') + mVal^{\sigma, \tau}(v'))$$

$\sigma$  Adam     $\tau$  Eve

Value

$$mVal(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$mVal^{\sigma}(v)$

Compute  $mVal^{\sigma}$

► If  $p < \frac{9}{10}$ , then  $q = 1$ :

$$mVal^{\sigma}(v_1) = \frac{-p}{1-p} > -9$$

$$mVal^{\sigma}(v_0) = \frac{-1}{1-p} > -10$$

► If  $p \geq \frac{9}{10}$ , then  $q = 0$ :

$$mVal^{\sigma}(v_1) = -10p$$

$$mVal^{\sigma}(v_0) = -10$$

# Memoryless strategies

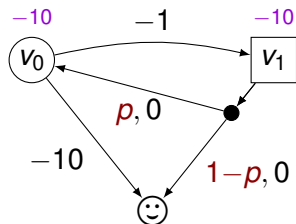
$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$

$\sigma$  Adam       $\tau$  Eve

Value

$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$



## Value in a MDP

Computable in polynomial time

## Bellman equation in a Markov Chain

$$\text{mVal}^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v')(w(v, v') + \text{mVal}^{\sigma, \tau}(v'))$$

*Stochastic Shortest Paths and Weight-Bounded Properties in Markov Decision Processes*, C. Baier, N. Bertrand, C. Dubslaff, D. Gburek and O. Sankur, 2018, LICS.

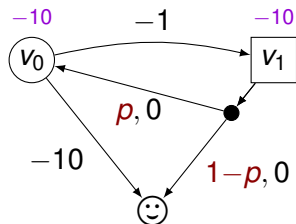


# Memoryless strategies

$$\sigma : V_{\text{Adam}} \rightarrow \Delta(V)$$

$\sigma$  Adam     $\tau$  Eve

Value



$$\text{mVal}(v) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_v^{\sigma, \tau}(\mathbf{SP})$$

$$\underbrace{\hspace{10em}}_{\text{mVal}^{\sigma}(v)}$$

## Value in a MDP

Computable in polynomial time

## $\epsilon$ -optimal strategy

$$\text{mVal}^{\sigma^*}(v) \leq \text{mVal}(v) + \epsilon$$

## Bellman equation in a Markov Chain

$$\text{mVal}^{\sigma, \tau}(v) = \sum_{v'} \mathbb{P}^{\sigma, \tau}(v, v')(w(v, v') + \text{mVal}^{\sigma, \tau}(v'))$$

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*Stochastic Shortest Paths and Weight-Bounded Properties in Markov Decision Processes*, C. Baier, N. Bertrand, C. Dubslaff, D. Gburek and O. Sankur, 2018, LICS.

## Contribution

$$\text{dVal} = \text{mVal}$$

# Contribution

Trade-off between memory and randomness

$$\text{dVal} = \text{mVal}$$

# Contribution

Trade-off between memory and randomness

- ▶ Stochastic games with qualitative objectives

$$\text{dVal} = \text{mVal}$$

# Contribution

## Trade-off between memory and randomness

- ▶ Stochastic games with qualitative objectives
- ▶ Reachability Timed Games

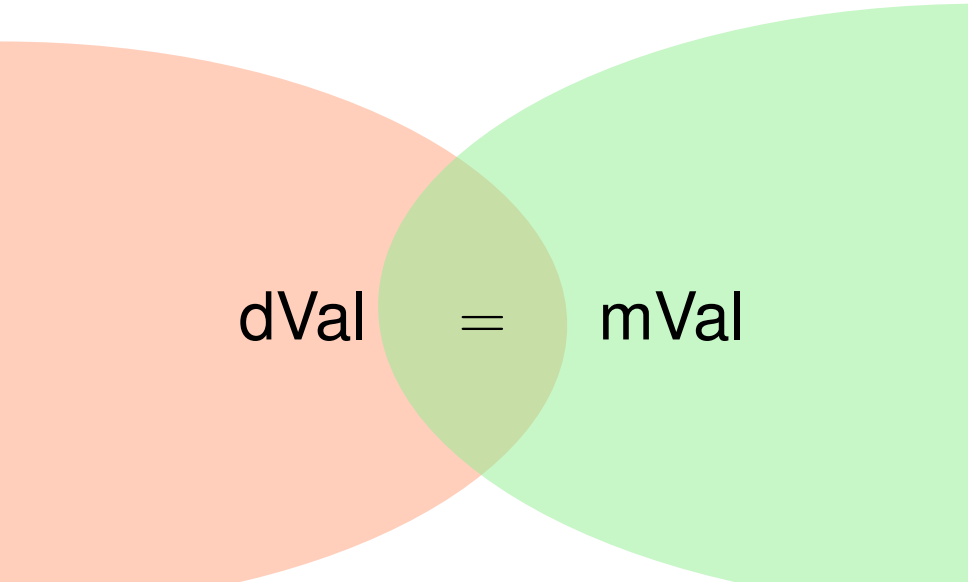
$$\text{dVal} = \text{mVal}$$

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*Trading Memory for Randomness*, K. Chatterjee, L. Alfaró and T. Henzinger, 2004, QEST

*Trading Infinite Memory for Uniform Randomness in Timed Games*, K. Chatterjee, T. Henzinger and S. Vinayak, 2008, HSCC

# Contribution



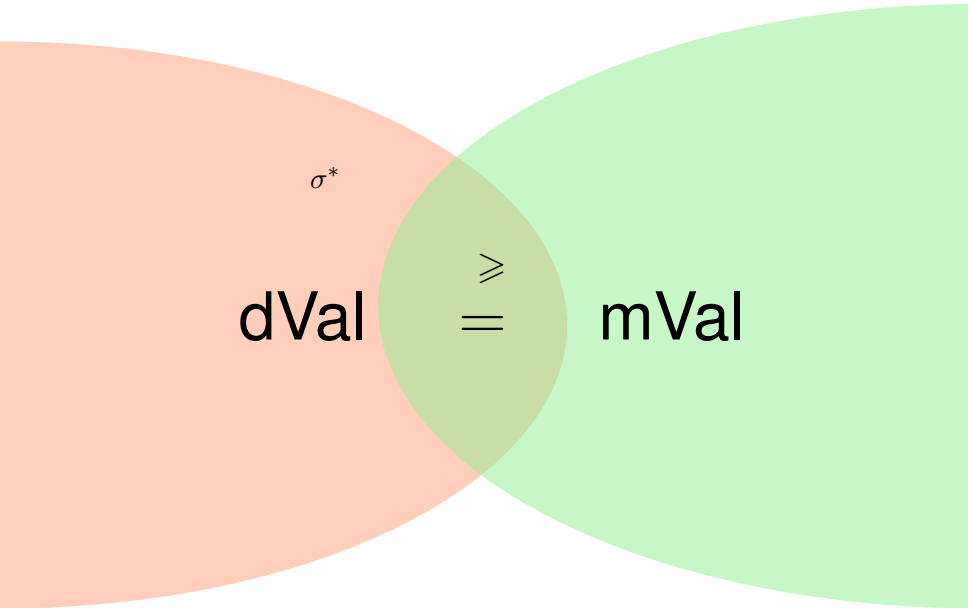
A Venn diagram consisting of two overlapping circles. The left circle is light orange and contains the text 'dVal'. The right circle is light green and contains the text 'mVal'. The overlapping area in the center is a darker shade of green and contains an equals sign '='. The text 'dVal = mVal' is centered across the diagram.

dVal = mVal

# Contribution



# Contribution





# Contribution

Switching  
strategy



dVal

$\geq$   
 $=$

mVal

# Contribution

Switching  
strategy



dVal

$\geq$   
 $=$

mVal

$\rho p$

# Contribution

Switching  
strategy



dVal

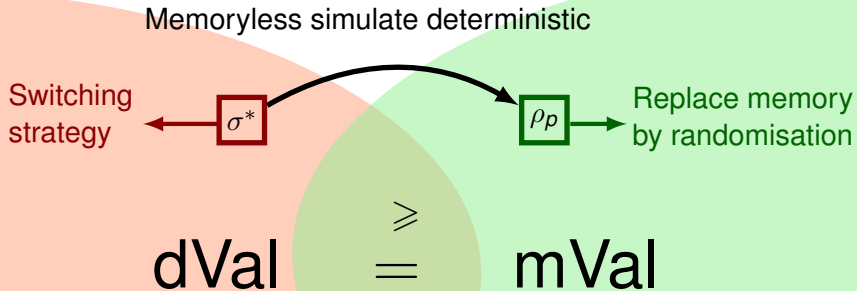
$\geq$   
 $=$



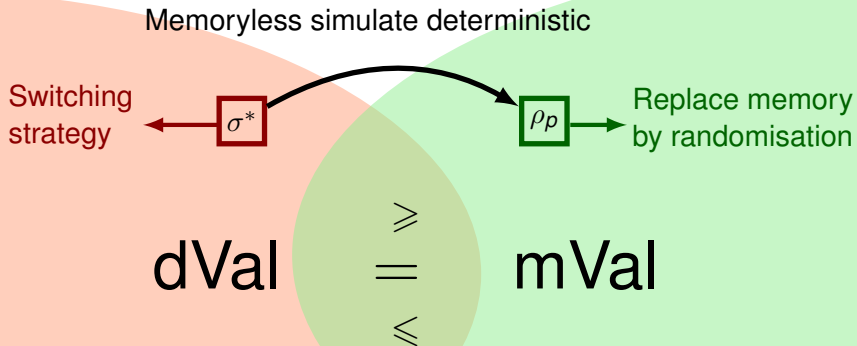
Replace memory  
by randomisation

mVal

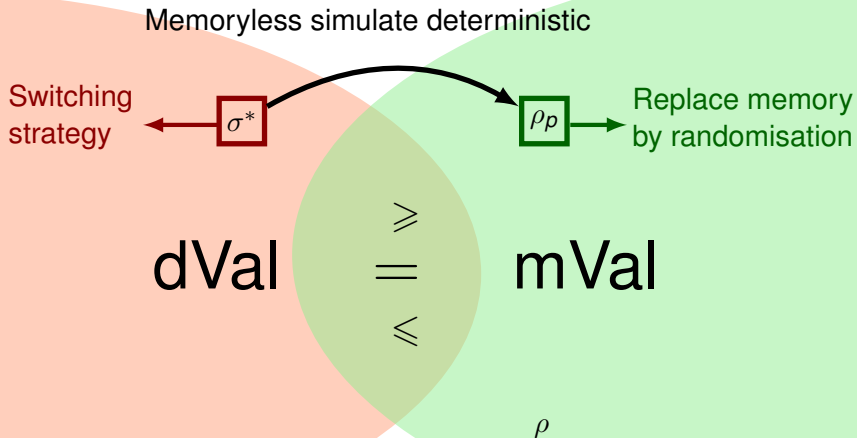
# Contribution



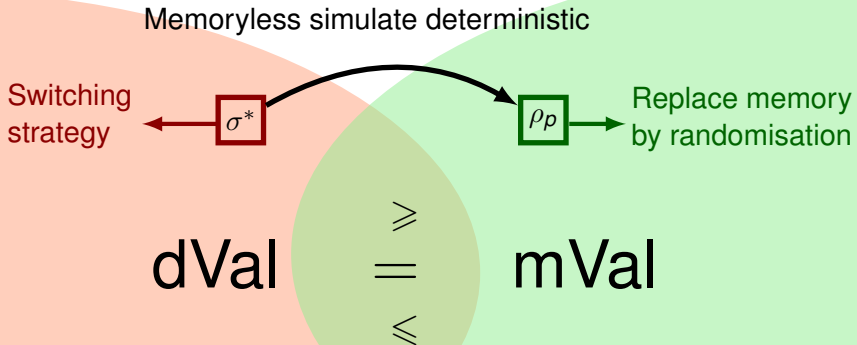
# Contribution



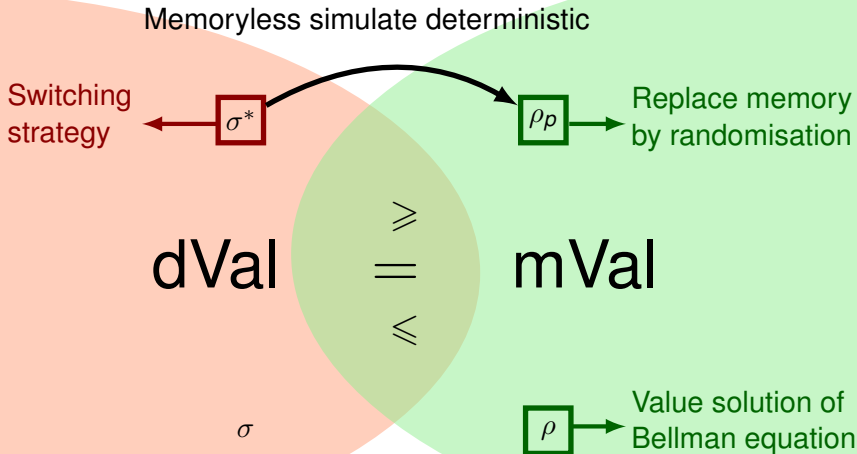
# Contribution



# Contribution

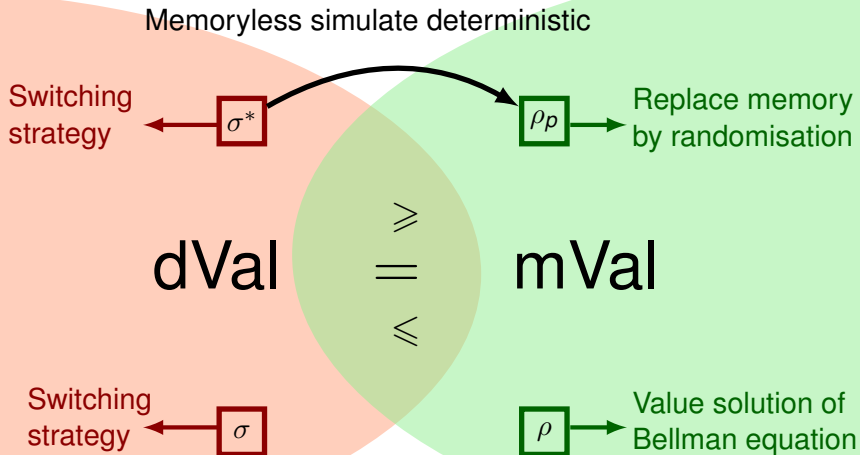


# Contribution

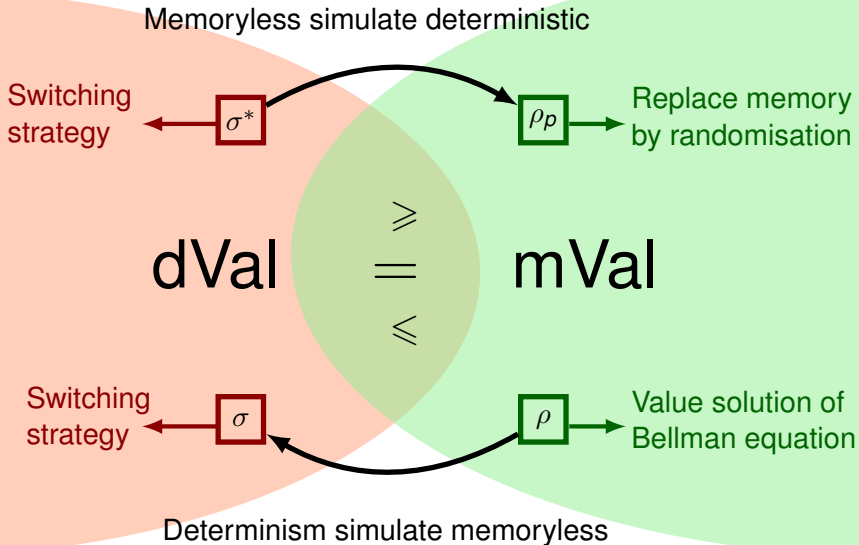




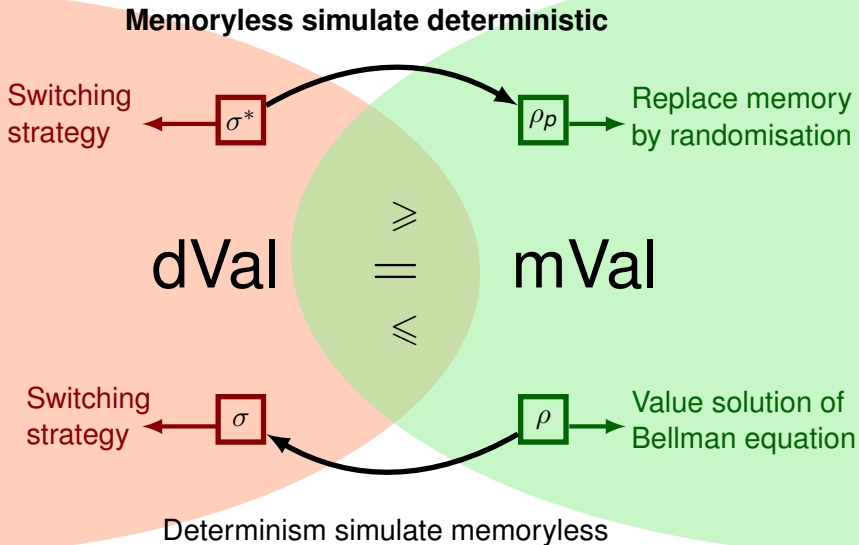
# Contribution



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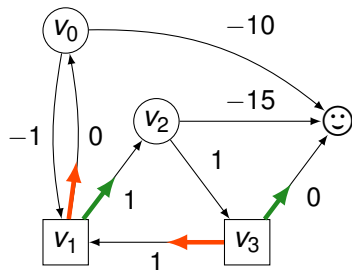
# Memoryless simulate deterministic

## Claim

For all  $v$ , there exists  $\rho$  such that  $mVal^{\rho\rho}(v) \leq dVal(v)$ .

# Memoryless simulate deterministic

□ Adam    ○ Eve



## Claim

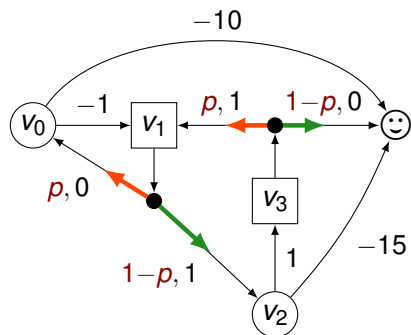
For all  $v$ , there exists  $p$  such that  $mVal^{\rho p}(v) \leq dVal(v)$ .

## Strategy $\rho_p$

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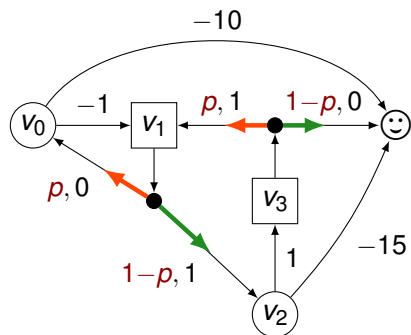
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- ▶ For all  $\tau$ ,  $\mathbb{P}^{\rho_p, \tau}(\diamond \text{smiley}) = 1$

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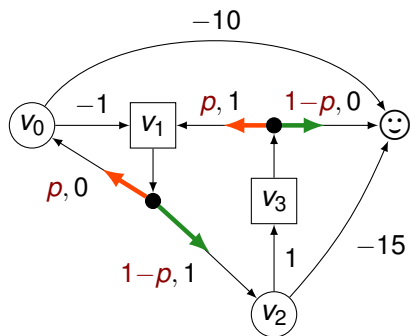
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- ▶ Eve has an optimal memoryless deterministic strategy.

## Strategy $\rho_p$

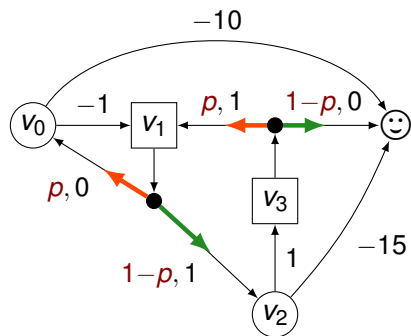
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Presence of non-negative cycles

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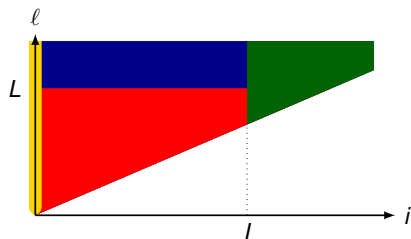
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Presence of non-negative cycles

## Tool for the proof

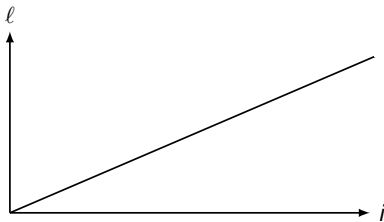
Control the non-negative cycles with a partition of plays

## Strategy $\rho_p$

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Focus on the partition of plays  $\ell$  size of play reaching the target  
 $i$  number of non-negative cycles

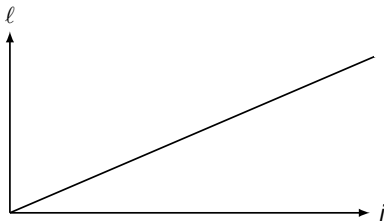


# Focus on the partition of plays

Fix a deterministic strategy for Eve

$\ell$  size of play reaching the target

$i$  number of non-negative cycles



# Focus on the partition of plays

$\ell$  size of play reaching the target

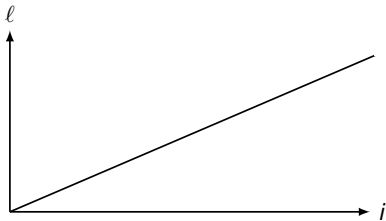
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Fix a deterministic strategy for Eve

Good zones

$$\mathbf{SP} \leq d\text{Val}$$

$$\Rightarrow \mathbb{E}(\mathbf{SP}) \leq d\text{Val}$$



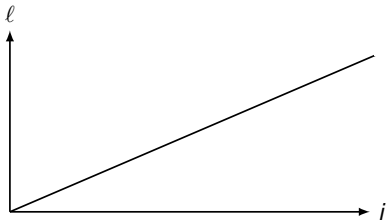
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Zones to control

$$\mathbb{E}(\mathbf{SP}) \leq \varepsilon$$

# Focus on the partition of plays

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Fix a deterministic strategy for Eve

## Yellow zone

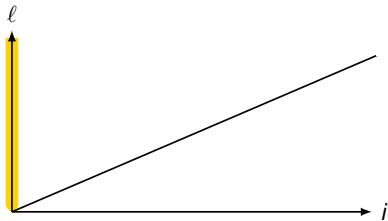
All plays conforming to  $\sigma_1$



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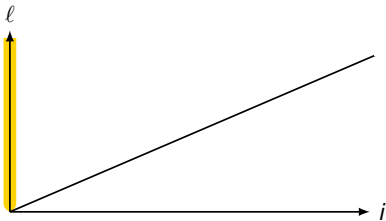
Weight of each play is  $\leq dVal$



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## Zones to control

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Fix a deterministic strategy for Eve

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## Yellow zone

All plays conforming to  $\sigma_1$

Weight of each play is  $\leq dVal$

## Green zone

Plays contain many non-negative cycles

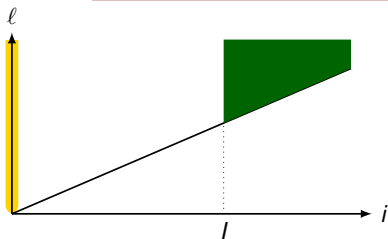
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$\forall p, \exists l$  s.t. expectation  $\leq \frac{\epsilon}{2}$

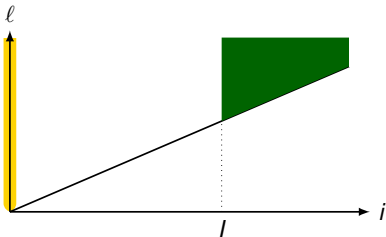
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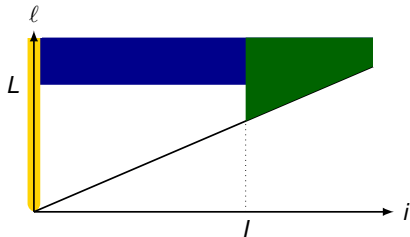
## Good zones

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## Blue zone

Plays with many negative cycles and few non-negative cycles



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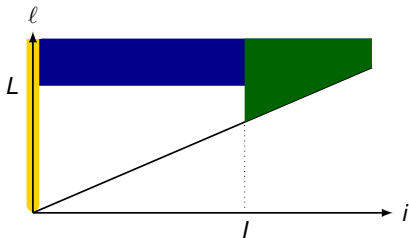
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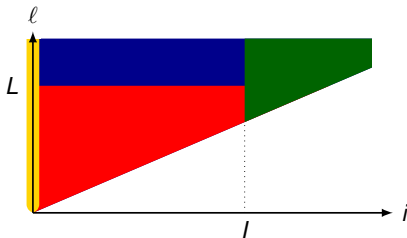
Plays contain many non-negative cycles

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## Blue zone

Plays with many negative cycles and few non-negative cycles

$\forall p, l, \exists L$  s.t. weights  $\leq dVal$

## Red zone

Rest of plays

# Focus on the partition of plays

$\ell$  size of play reaching the target

$i$  number of non-negative cycles

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## Yellow zone

All plays conforming to  $\sigma_1$

Weight of each play is  $\leq dVal$

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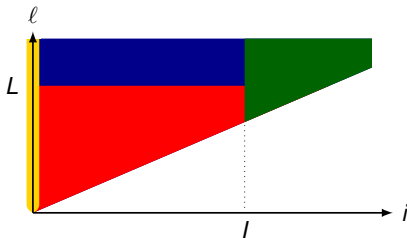
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Rest of plays

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# Results on shortest path games

## Contributions

1. Adam has the same hope using memory or randomness.
2. Existence of an optimal memoryless strategy for Adam is testable in polynomial time.

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- ▶ Extension to probabilistic value (memory and randomisation)
- ▶ Extension to weighted timed games
- ▶ A polynomial-time algorithm to compute the value



# Results on shortest path games

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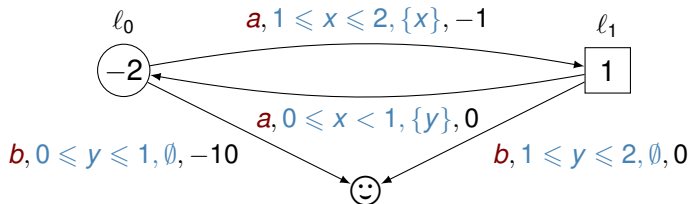
1. Adam has the same hope using memory or randomness.
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## Perspectives

- ▶ Extension to probabilistic value (memory and randomisation)
- ▶ **Extension to weighted timed games**
- ▶ A polynomial-time algorithm to compute the value

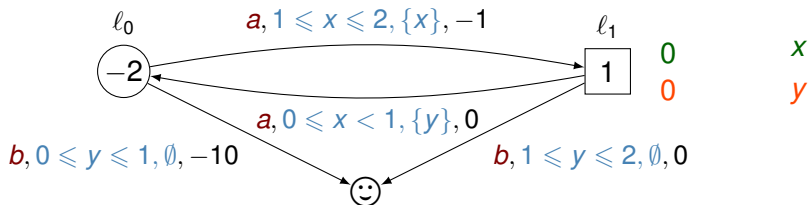
# Weighted timed games

□ Adam    ○ Eve



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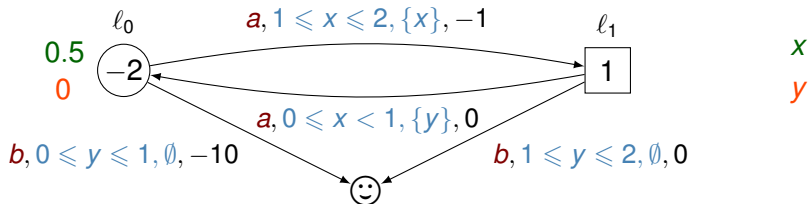


Play

$(l_1, \begin{pmatrix} 0 \\ 0 \end{pmatrix})$

# Weighted timed games

□ Adam    ○ Eve



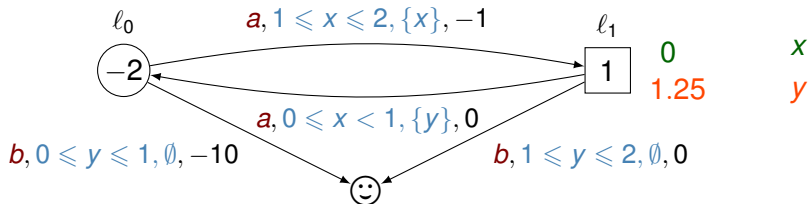
## Play

$$\left( l_1, \begin{pmatrix} 0 \\ 0 \end{pmatrix} \right) \xrightarrow{a, 0.5} \left( l_0, \begin{pmatrix} 0.5 \\ 0 \end{pmatrix} \right)$$

$1 \times 0.5 + 0$

# Weighted timed games

□ Adam    ○ Eve

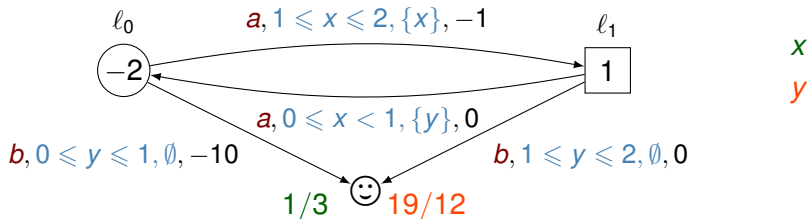


## Play

$$\begin{array}{c}
 (l_1, \begin{pmatrix} 0 \\ 0 \end{pmatrix}) \xrightarrow{a, 0.5} (l_0, \begin{pmatrix} 0.5 \\ 0 \end{pmatrix}) \xrightarrow{a, 1.25} (l_1, \begin{pmatrix} 0 \\ 1.25 \end{pmatrix}) \\
 1 \times 0.5 + 0 \qquad -2 \times 1.25 - 1
 \end{array}$$

# Weighted timed games

□ Adam    ○ Eve

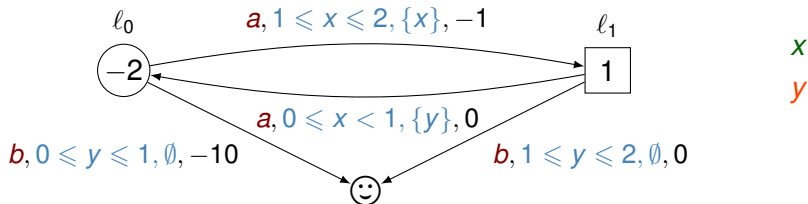


## Play

$$\begin{aligned}
 (l_1, \begin{pmatrix} 0 \\ 0 \end{pmatrix}) &\xrightarrow{a, 0.5} (l_0, \begin{pmatrix} 0.5 \\ 0 \end{pmatrix}) \xrightarrow{a, 1.25} (l_1, \begin{pmatrix} 0 \\ 1.25 \end{pmatrix}) \xrightarrow{b, 1/3} (\text{smiley face}, \begin{pmatrix} 1/3 \\ 19/12 \end{pmatrix}) \\
 1 \times 0.5 + 0 & \quad -2 \times 1.25 - 1 & \quad 1 \times \frac{1}{3} + 0
 \end{aligned}$$

# Weighted timed games

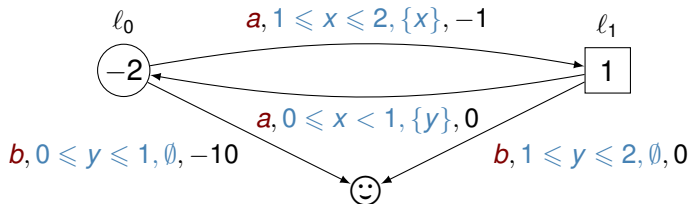
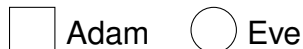
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 &\quad 1 \times 0.5 + 0 \qquad -2 \times 1.25 - 1 \qquad 1 \times \frac{1}{3} + 0
 \end{aligned}$$

# Weighted timed games



## Value problem

The value problem, i.e. deciding if  $dVal(l_0, \nu_0) \leq c$  with  $c \in \mathbb{Z}$ , is undecidable.

---

*On Optimal Timed Strategies*, T.Brihaye, V.Bruyère and J.-F. Raskin, 2005, FORMATS.

*On the value problem in weighted timed games*, P. Bouyer, S. Jaziri, and N. Markey, 2015, CONCUR.




## Our objective

dVal  $\stackrel{?}{=}$  mVal

## Our objective

Define probabilistic strategy

dVal  $\stackrel{?}{=}$  mVal

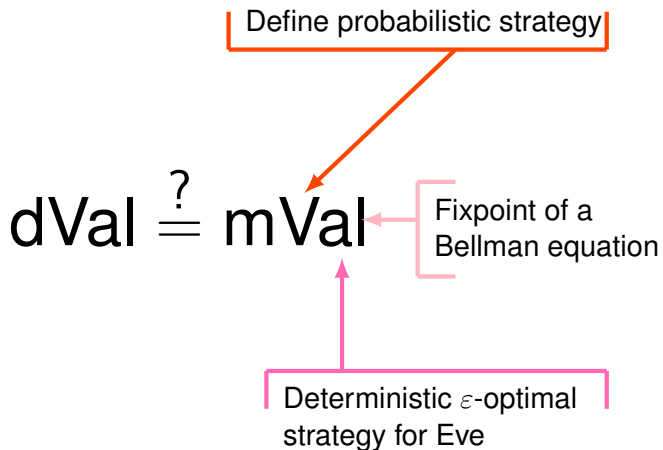


## Our objective

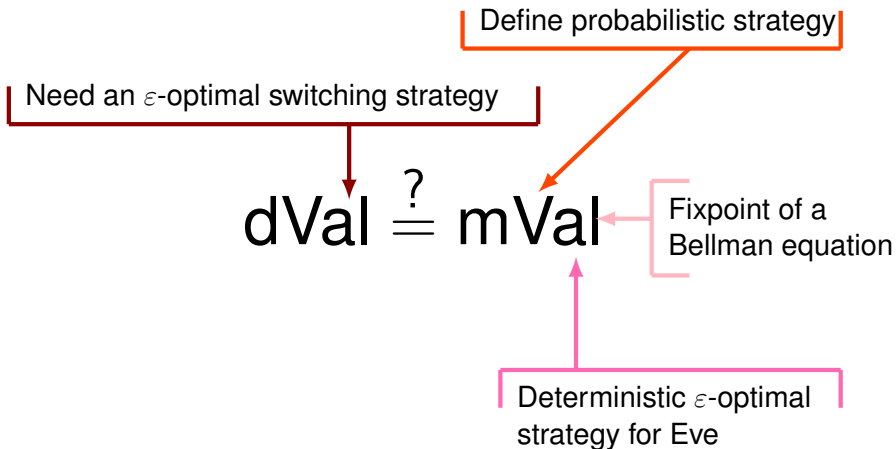
Define probabilistic strategy

$dVal \stackrel{?}{=} mVal$  Fixpoint of a  
Bellman equation

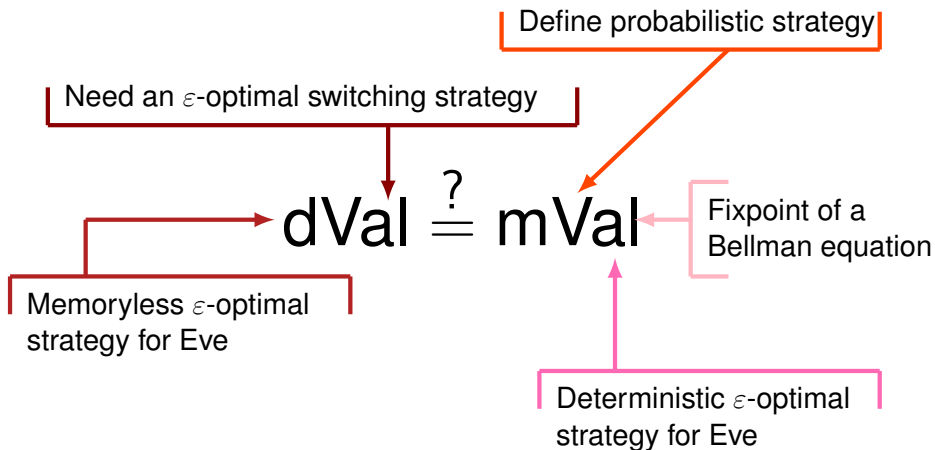
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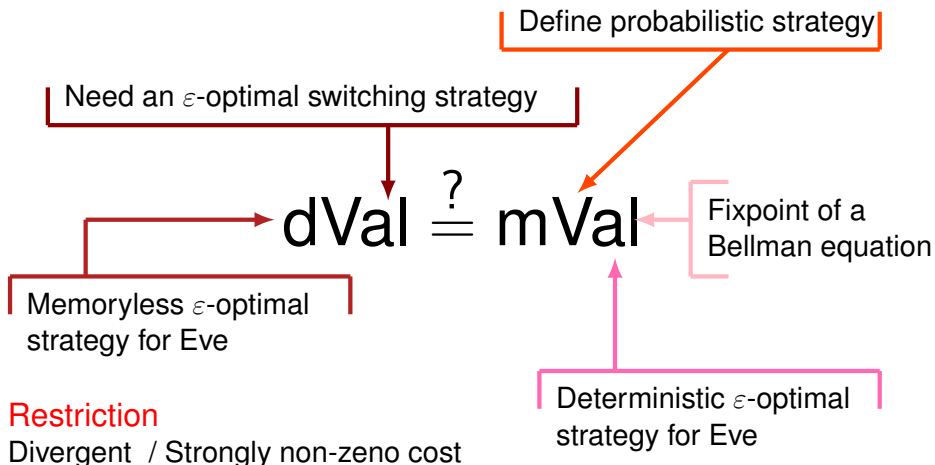
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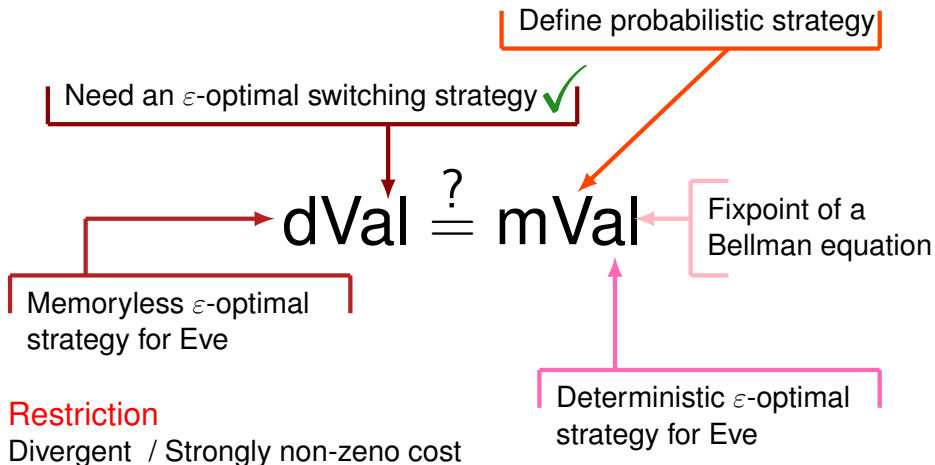
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*Optimal Reachability in Divergent Weighted Timed Games*, D. Busatto-Gaston, B. Monmege and P.-A. Reynier, 2017, ETAPS

*Optimal Strategies in Priced Timed Game Automata*, P. Bouyer, F. Cassez, E. Fleury and K. Larsen, 2004, FSTTCS

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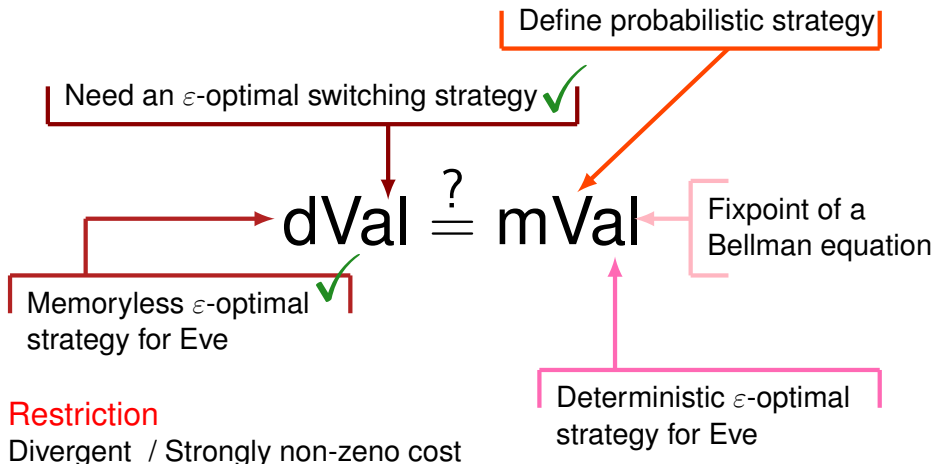


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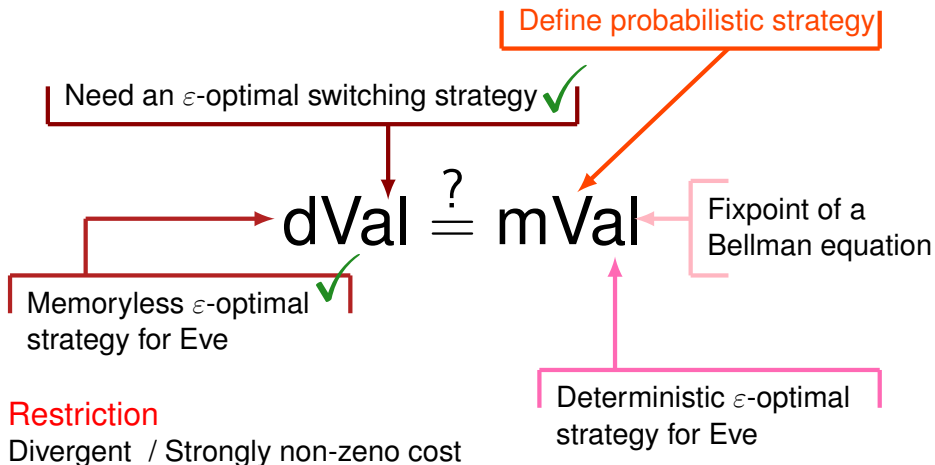
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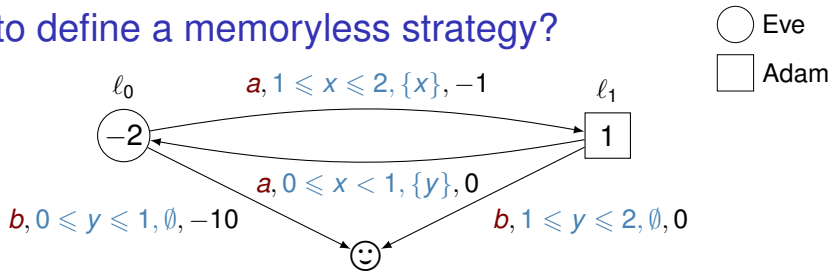
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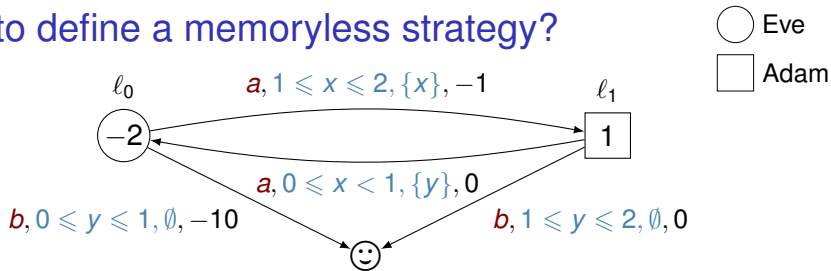
# How to define a memoryless strategy?



## Deterministic strategy

Choose a transition and a delay

# How to define a memoryless strategy?



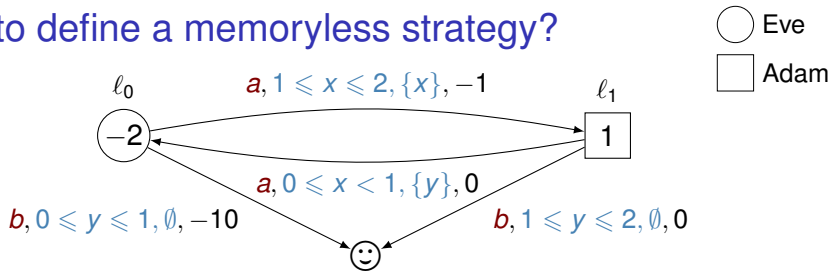
## Deterministic strategy

Choose a transition and a delay

In  $(l_1, (0, 0))$

Choose between  $a$  or  $b$

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## Deterministic strategy

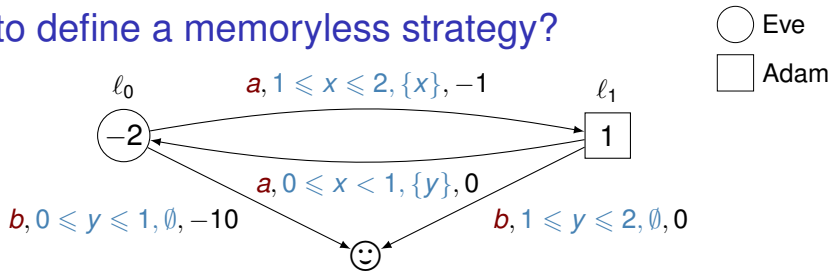
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- ▶  $a$ : choose  $t$  with  $t < 1$

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## Deterministic strategy

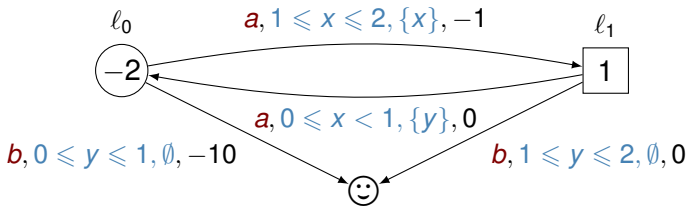
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In  $(l_1, (0, 0))$

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- ▶  $b$ : choose  $t$  with  $1 \leq t \leq 2$

# How to define a memoryless strategy?



## Probabilistic strategy

Distribution over possible choices

## Deterministic strategy

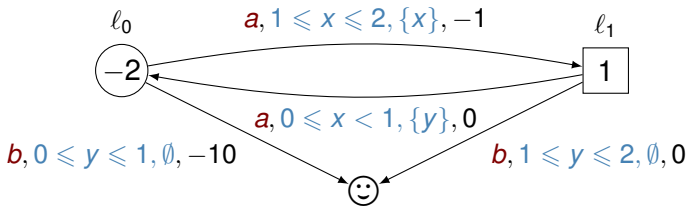
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# How to define a memoryless strategy?



## Probabilistic strategy

Distribution over possible choices

1. Transition  $a$ : finite distribution

## Deterministic strategy

Choose a transition and a delay

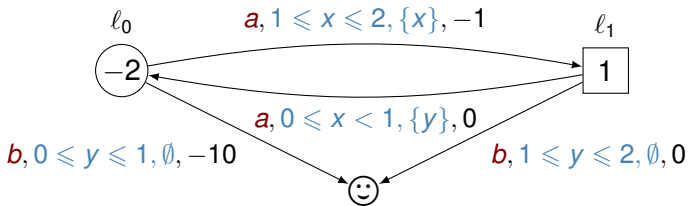
In  $(l_1, (0, 0))$

Choose between  $a$  or  $b$

- ▶  $a$ : choose  $t$  with  $t < 1$
- ▶  $b$ : choose  $t$  with  $1 \leq t \leq 2$



# How to define a memoryless strategy?



## Probabilistic strategy

Distribution over possible choices

1. Transition  $a$ : finite distribution
2. Delay for  $a$ : infinite distribution

## Deterministic strategy

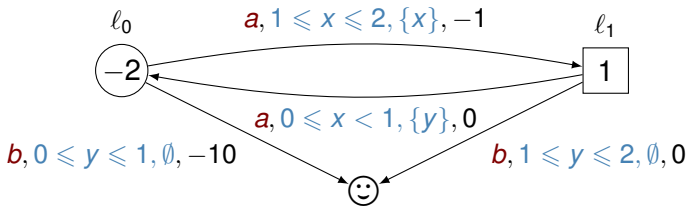
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Choose between  $a$  or  $b$  with  $\mathcal{B}(p)$

## Deterministic strategy

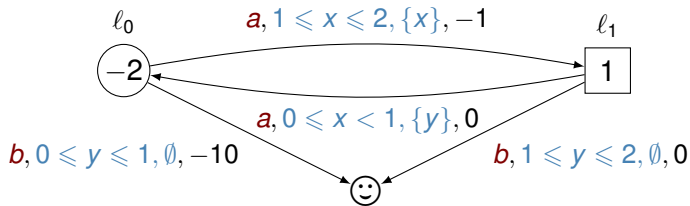
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- ▶  $a$ : choose  $t$  with  $\mathcal{U}([0, 1])$

## Deterministic strategy

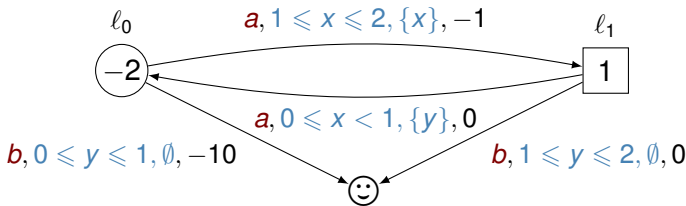
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- ▶  $a$ : choose  $t$  with  $\mathcal{U}([0, 1])$
- ▶  $b$ : choose  $t$  with  $\delta_{1.5}$

## Deterministic strategy

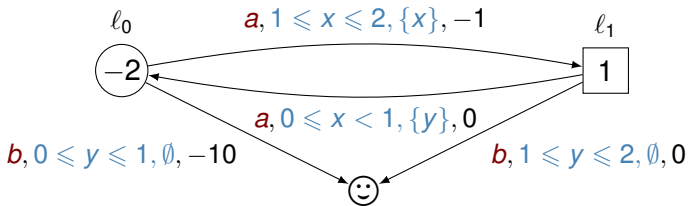
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# How to define a memoryless strategy?



## Probabilistic strategy

Distribution over possible choices

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2. Delay for  $a$ : infinite distribution

## Deterministic strategy

Choose a transition and a delay

## Close related work

Stochastic timed automata

## Memoryless value

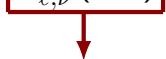
$$\overline{\text{mVal}}(\ell, \nu) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_{\ell, \nu}^{\sigma, \tau}(\mathbf{SP})$$

## Memoryless value

$$\overline{\text{mVal}}(\ell, \nu) = \inf_{\sigma} \sup_{\tau} \underbrace{\mathbb{E}_{\ell, \nu}^{\sigma, \tau}(\mathbf{SP})}_{\downarrow}$$
$$\sum_{\mathcal{C}} \mathbb{P}(\mathcal{C}) \mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C})$$


# Memoryless value

Cylinder

$$\overline{\text{mVal}}(\ell, \nu) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_{\ell, \nu}^{\sigma, \tau}(\mathbf{SP})$$


$$\mathcal{C} = ((\ell, \nu), \mathbf{e}_1 \dots \mathbf{e}_n) =$$

$$\{(t_1, \dots, t_n) \mid (\ell, \nu) \xrightarrow{t_1, \mathbf{e}_1} \dots \xrightarrow{t_n, \mathbf{e}_n}\}$$

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# Memoryless value

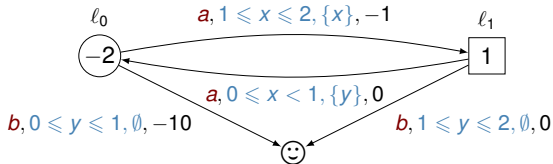


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Example of cylinder  $\mathcal{C}$

$$\mathcal{C} = ((l_1, (0, 0)), \mathbf{a} \mathbf{b}) =$$

# Memoryless value



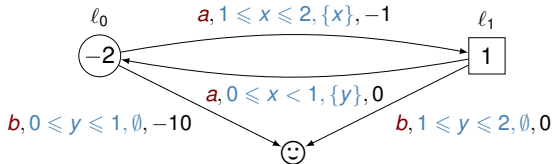
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# Memoryless value

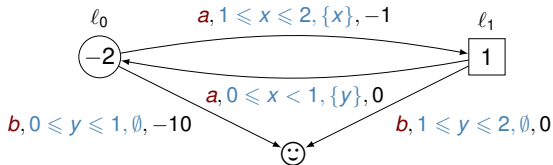


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## Cylinder

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## Example of cylinder $\mathcal{C}$

$$\begin{aligned} \mathcal{C} &= ((l_1, (0, 0)), \mathbf{a} \mathbf{b}) = \{(t_1, t_2) \mid (l_1, (0, 0)) \xrightarrow{t_1, \mathbf{a}} \xrightarrow{t_2, \mathbf{b}}\} \\ &= \{(l_1, \nu_0) \xrightarrow{t_1, \mathbf{a}} (l_0, \nu_1) \xrightarrow{t_2, \mathbf{b}} (\odot, \nu_2) \mid t_1 < 1, t_2 \leq 1\} \end{aligned}$$

Annotations for the cylinder example:

- $(0, 0)$  (orange) is reached from  $(l_1, \nu_0)$  via action  $\mathbf{a}$ .
- $(t_1, 0)$  (green) is reached from  $(l_0, \nu_1)$  via action  $\mathbf{a}$ .
- $(t_1 + t_2, t_2)$  (pink) is reached from  $(\odot, \nu_2)$  via action  $\mathbf{b}$ .

# Memoryless value

## Cylinder

$$\mathcal{C} = ((\ell, \nu), \mathbf{e}_1 \dots \mathbf{e}_n) = \{(t_1, \dots, t_n) \mid (\ell, \nu) \xrightarrow{t_1, \mathbf{e}_1} \dots \xrightarrow{t_n, \mathbf{e}_n}\}$$

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## Probability of a cylinder

$$\mathbb{P}(\mathcal{C}) = \int_{t_1 \in I((\ell, \nu), \mathbf{e}_1)} p_{(\ell, \nu)}(\mathbf{e}_1) \mathbb{P}(\mathcal{C}_1) d\mu_{(\ell, \nu), \mathbf{e}_1}(t_1)$$

# Memoryless value

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- Probability to choose  $\mathbf{e}_1$  in  $(\ell, \nu)$

# Memoryless value

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- ▶ Interval of possible delays to pass through  $\mathbf{e}_1$  from  $(\ell, \nu)$

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$$((\ell_1, \nu_1), \mathbf{e}_2 \dots \mathbf{e}_n)$$



# Memoryless value

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- ▶  $(\ell, \nu) \xrightarrow{t_1, \mathbf{e}_1} (\ell_1, \nu_1)$
- ▶  $n$ -dimensional integral

# Memoryless value

Cylinder

$$\overline{\text{mVal}}(\ell, \nu) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_{\ell, \nu}^{\sigma, \tau}(\mathbf{SP})$$

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Probability of a cylinder

$$\mathbb{P}(\mathcal{C}) = \int_{t_1} p_{(\ell, \nu)}(\mathbf{e}_1) \left( \int_{t_2} p_{(\ell_1, \nu_1)}(\mathbf{e}_2) \int_{t_3} \dots \right) d\mu_{(\ell, \nu), \mathbf{e}_1}(t_1)$$

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$$\sum_{\mathcal{C}} \mathbb{P}(\mathcal{C}) \mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C})$$

Conditional expectation

$$\mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}) = \int_{t_1} p_{(\ell, \nu)}(\mathbf{e}_1)(w(\ell)t_1 + w(\mathbf{e}_1) + \mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}_1)) d\mu_{(\ell, \nu), \mathbf{e}_1}(t_1)$$

Probability of a cylinder

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# Memoryless value

$$\overline{\text{mVal}}(\ell, \nu) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_{\ell, \nu}^{\sigma, \tau}(\mathbf{SP})$$

Cylinder

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Existence and integrability ?

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# Memoryless value

$$\overline{\text{mVal}}(\ell, \nu) = \inf_{\sigma} \sup_{\tau} \mathbb{E}_{\ell, \nu}^{\sigma, \tau}(\mathbf{SP})$$

Cylinder

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Existence and integrability ?

No restriction on Eve's strategies

Conditional expectation

$$\mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}) = \int_{t_1} p_{(\ell, \nu)}(\mathbf{e}_1)(w(\ell)t_1 + w(\mathbf{e}_1) + \mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}_1)) d\mu_{(\ell, \nu), \mathbf{e}_1}(t_1)$$

Probability of a cylinder

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# Memoryless value

## Cylinder

$$\mathcal{C} = ((\ell, \nu), \mathbf{e}_1 \dots \mathbf{e}_n) = \{(t_1, \dots, t_n) \mid (\ell, \nu) \xrightarrow{t_1, \mathbf{e}_1} \dots \xrightarrow{t_n, \mathbf{e}_n}\}$$

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## Existence and integrability ?

No restriction on Eve's strategies

Restriction on Adam's strategies

## Conditional expectation

$$\mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}) = \int_{t_1} p_{(\ell, \nu)}(\mathbf{e}_1)(w(\ell)t_1 + w(\mathbf{e}_1) + \mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}_1)) d\mu_{(\ell, \nu), \mathbf{e}_1}(t_1)$$

## Probability of a cylinder

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# Memoryless value

## Cylinder

$$\mathcal{C} = ((\ell, \nu), \mathbf{e}_1 \dots \mathbf{e}_n) = \{(t_1, \dots, t_n) \mid (\ell, \nu) \xrightarrow{t_1, \mathbf{e}_1} \dots \xrightarrow{t_n, \mathbf{e}_n}\}$$

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## Existence and integrability ?

No restriction on Eve's strategies  
Restriction on Adam's strategies

## Convergence ?

- ▶  $\mathbb{P}(\mathcal{C}) \sim \alpha^{-n}$
- ▶  $\mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}) \sim kn$

## Conditional expectation

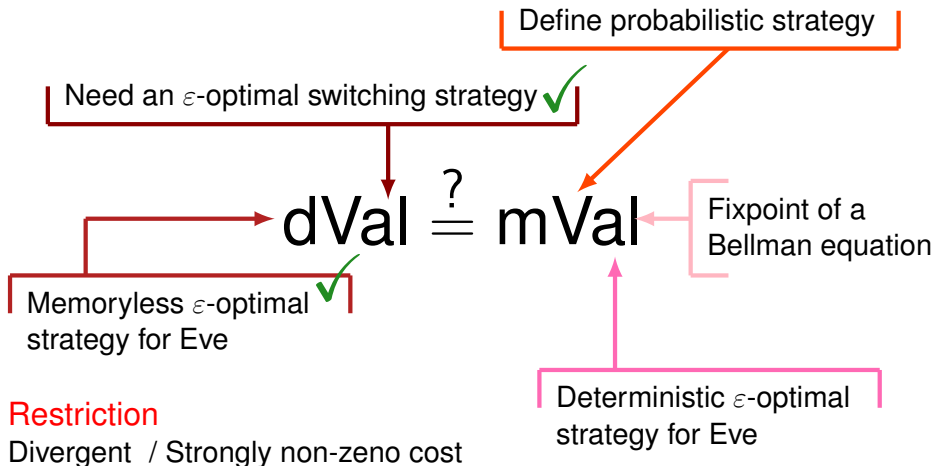
$$\mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}) = \int_{t_1} p_{(\ell, \nu)}(\mathbf{e}_1)(w(\ell)t_1 + w(\mathbf{e}_1) + \mathbb{E}^{\sigma, \tau}(\mathbf{SP} \mid \mathcal{C}_1)) d\mu_{(\ell, \nu), \mathbf{e}_1}(t_1)$$

## Probability of a cylinder

$$\mathbb{P}(\mathcal{C}) = \int_{t_1} p_{(\ell, \nu)}(\mathbf{e}_1) \left( \int_{t_2} p_{(\ell_1, \nu_1)}(\mathbf{e}_2) \int_{t_3} \dots \right) d\mu_{(\ell, \nu), \mathbf{e}_1}(t_1) \mathbb{P}(\mathcal{C}_1)$$



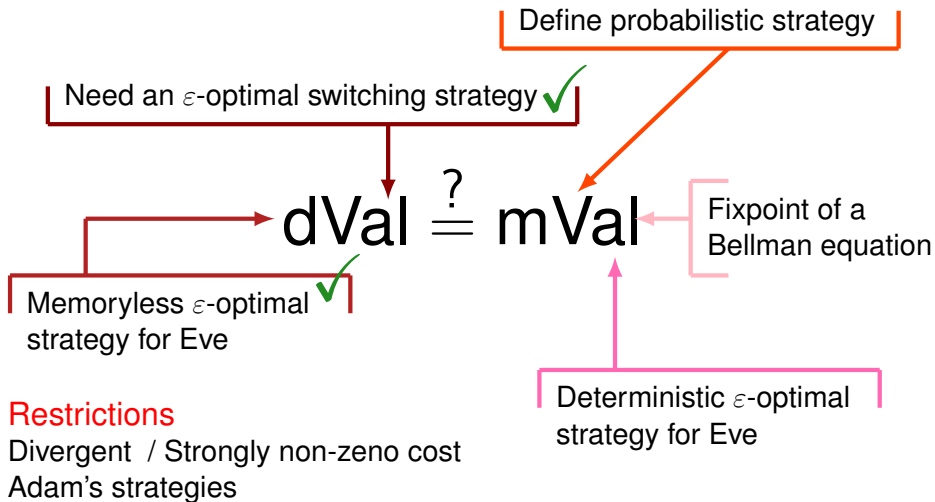
# To conclude



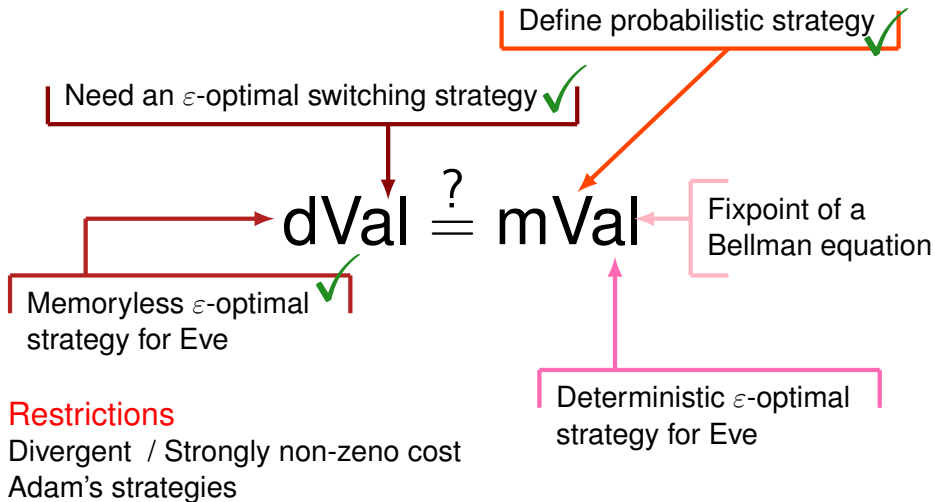
*Optimal Reachability in Divergent Weighted Timed Games*, D. Busatto-Gaston, B. Monmege and P.-A. Reynier, 2017, ETAPS

*Optimal Strategies in Priced Timed Game Automata*, P. Bouyer, F. Cassez, E. Fleury and K. Larsen, 2004, FSTTCS

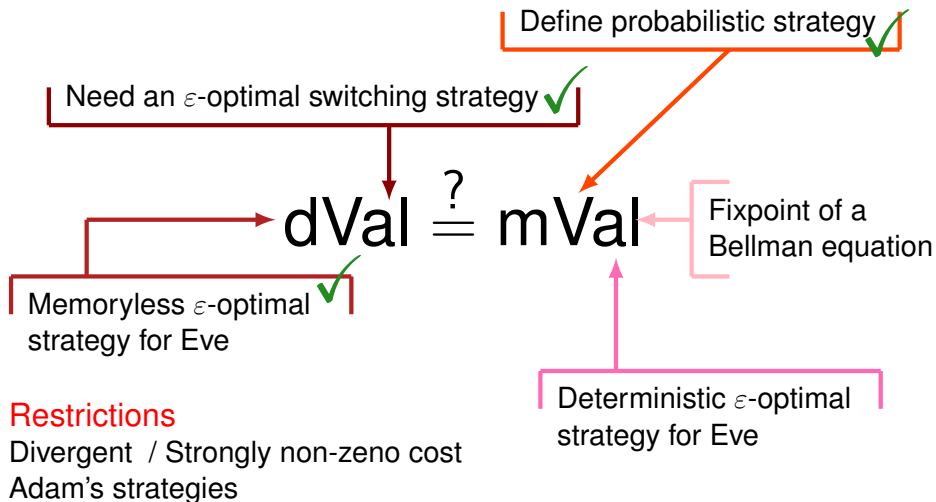
## To conclude



## To conclude



## To conclude



Thank you! Questions?