

# Multi-Weighted Reachability Games

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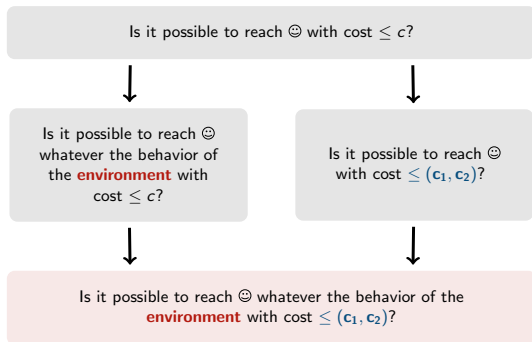
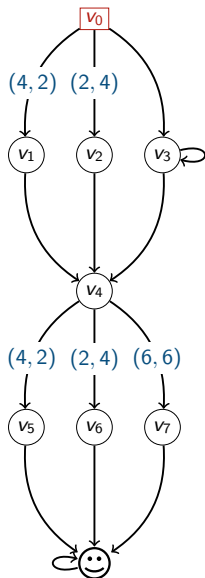
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2 Studied problems

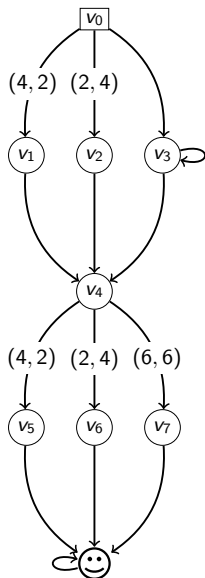
3 Conclusion

# Reachability Games



## Two-Player Multi-Weighted Reachability Games

# Two-Player Multi-Weighted Reachability Games



- Turn-based;
- Two players: Player  $\bigcirc$  and Player  $\square$ ;
- A  $d$ -weighted graph  $G = (V, E, (w_i)_{1 \leq i \leq d})$

## Quantitative reachability objective

Given a target set  $F \subseteq V$ , for all **plays** (infinite paths in  $G$ )  $\rho = \rho_0 \rho_1 \dots$ :

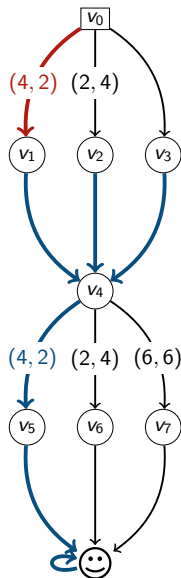
$$\text{Cost}_i(\rho) = \begin{cases} \sum_{n=0}^{k-1} w_i(\rho_n, \rho_{n+1}) & \text{if } k \text{ is the least} \\ & \text{index st. } \rho_k \in F \\ +\infty & \text{otherwise} \end{cases}$$

Rem: **same target set** for all dimensions.

Ex:

- $\text{Cost}(v_0 v_3^\omega) = (\text{Cost}_1(v_0 v_3^\omega), \text{Cost}_2(v_0 v_3^\omega)) = (+\infty, +\infty)$
- $\text{Cost}(v_0 v_1 v_4 v_5 (\odot)^\omega) = (10, 6)$

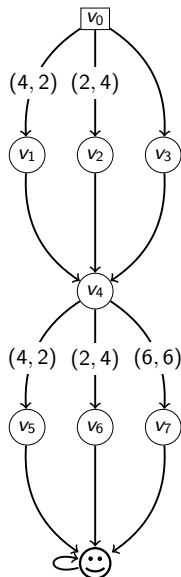
# Two-Player Multi-Weighted Reachability Games



- A strategy for Player  $\bigcirc$ :  $\sigma_{\bigcirc} : V^* V_{\bigcirc} \longrightarrow V$ .
- Given a **strategy profile**  $(\sigma_{\bigcirc}, \sigma_{\square})$  and an initial vertex  $v_0 \rightsquigarrow$  only one consistent play  $\langle \sigma_{\bigcirc}, \sigma_{\square} \rangle_{v_0}$  called the **outcome**.

Ex:  $\langle \sigma_{\bigcirc}, \sigma_{\square} \rangle_{v_0} = v_0 v_1 v_4 v_5 (\text{smiley})^{\omega}$ .

# Two-Player Multi-Weighted Reachability Games



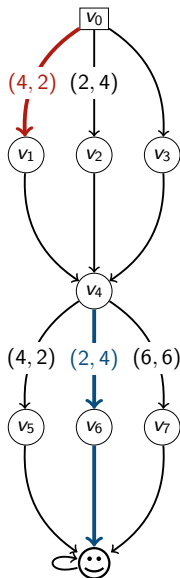
Player  $\bigcirc$  can **ensure** a cost profile  $\mathbf{c} = (c_1, \dots, c_d)$  from  $v$  if **there exists** a strategy  $\sigma_{\bigcirc}$  such that **for all strategies**  $\sigma_{\square}$  of Player  $\square$ :

$$\text{Cost}(\langle \sigma_{\bigcirc}, \sigma_{\square} \rangle_v) \leq_{\mathbf{c}} \mathbf{c} = (c_1, \dots, c_d)$$

Ex:

- $(8, 8) \rightsquigarrow$  **Yes.** (with memory!)

# Two-Player Multi-Weighted Reachability Games



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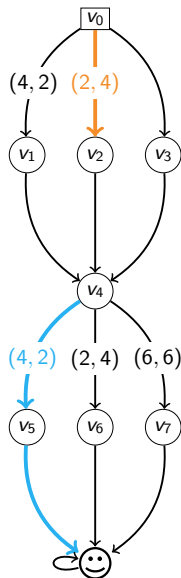
■  $(8, 8) \rightsquigarrow$  **Yes. (with memory!)**

■  $\mapsto$  ■  $(8, 8)$

Player  $\bigcirc$  can adapt his strategy in function of the choice of Player  $\square \rightsquigarrow$  finite-memory strategy!



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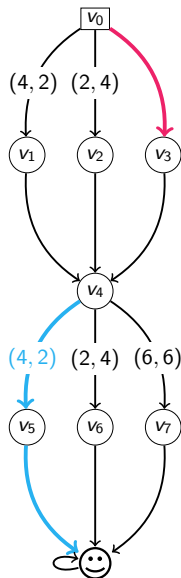
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# Two-Player Multi-Weighted Reachability Games









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$$\text{Cost}(\langle \sigma_{\bigcirc}, \sigma_{\square} \rangle_v) \leq_{\mathbf{c}} \mathbf{c} = (c_1, \dots, c_d)$$

Ex:

■  $(8, 8) \rightsquigarrow$  **Yes. (with memory!)**

	$\mapsto$		$(8, 8)$
	$\mapsto$		$(8, 8)$
	$\mapsto$		$(7, 5)$

Player  $\bigcirc$  can adapt his strategy in function of the choice of Player  $\square \rightsquigarrow$  finite-memory strategy!

$$\text{Ensure}(v) = \{\mathbf{c} \in \overline{\mathbb{N}}^d \mid \exists \sigma_{\bigcirc} \text{ st. } \forall \sigma_{\square}, \text{Cost}(\langle \sigma_{\bigcirc}, \sigma_{\square} \rangle_v) \leq_{\mathbf{c}} \mathbf{c}\}.$$

$\text{minimal}(\text{Ensure}(v)) = \text{Pareto}(v) \rightsquigarrow$  **Pareto frontier** from  $v$ .

For  $\mathbf{c} = (c_1, \dots, c_d) \in \text{Pareto}(v)$ , a strategy  $\sigma_{\bigcirc}$  is **c-Pareto-optimal** if  $\sigma_{\bigcirc}$  ensures  $\mathbf{c}$  from  $v$ .

Studied problems

- 1 Compute the Pareto frontier and Pareto-optimal strategies.
- 2 Decide the **constrained existence problem**.

### Constrained existence problem (CEP)

Given a game, a vertex  $v \in V$  and  $\mathbf{c} \in \mathbb{N}^d$ ,  
does there exist a strategy  $\sigma_{\bigcirc}$  of Player  $\bigcirc$  such that for all strategies of  
Player  $\square$ , we have:

$$\text{Cost}(\langle \sigma_{\bigcirc}, \sigma_{\square} \rangle_v) \leq_{\mathbf{c}} \mathbf{c}$$

## Computing Pareto( $v$ )

$$\text{Ensure}^k(v) = \{\mathbf{c} \in \overline{\mathbb{N}}^d \mid \exists \sigma_{\bigcirc} \text{ st. } \forall \sigma_{\square}, \\ \text{Cost}(\langle \sigma_{\bigcirc}, \sigma_{\square} \rangle_v) \leq_{\mathbf{c}} \mathbf{c} \wedge |\langle \sigma_{\bigcirc}, \sigma_{\square} \rangle_v|_F \leq k\}.$$

The algorithm computes, step by step, the sets  $I^k(v)$  for all  $v \in V$ .

For all  $k \in \mathbb{N}$  and all  $v \in V$ ,  $I^k(v) = \text{minimal}(\text{Ensure}^k(v))$

There exists  $k^* \in \mathbb{N}$  such that for all  $v \in V$  and for all  $\ell \in \mathbb{N}$ ,  
 $I^{k^*}(v) = I^{k^*+\ell}(v)$ .

For all  $v \in V$ ,  $I^{k^*}(v) = \text{Pareto}(v)$ .

### Theorem

The fixpoint algorithm runs in time polynomial in  $W$  and  $|V|$  and is **exponential** in  $d$ .

Where  $W$  is the maximal weight on an edge.

# Computing $\text{Pareto}(v)$ and Pareto-optimal strategies

```
for  $v \in F$  do  $I^0(v) = \{0\}$ 
for  $v \notin F$  do  $I^0(v) = \{\infty\}$ 

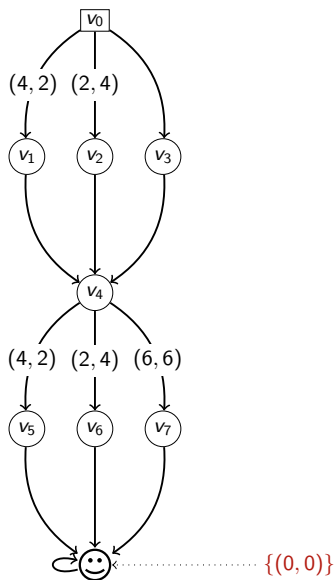
repeat
  for  $v \in V$  do
    if  $v \in F$  then  $I^{k+1}(v) = \{0\}$ 

    else if  $v \in V_{\square}$  then
      
$$I^{k+1}(v) = \text{minimal} \left( \bigcup_{v' \in \text{succ}(v)} \uparrow I^k(v') + \mathbf{w}(v, v') \right)$$


    else if  $v \in V_{\lozenge}$  then
      
$$I^{k+1}(v) = \text{minimal} \left( \bigcap_{v' \in \text{succ}(v)} \uparrow I^k(v') + \mathbf{w}(v, v') \right)$$

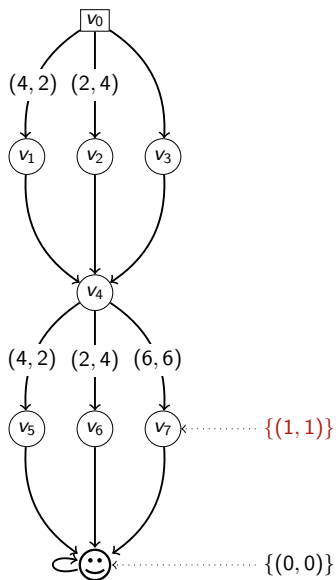

  until  $I^{k+1}(v) = I^k(v)$  for all  $v \in V$ 
```

## Computing Pareto( $v$ )

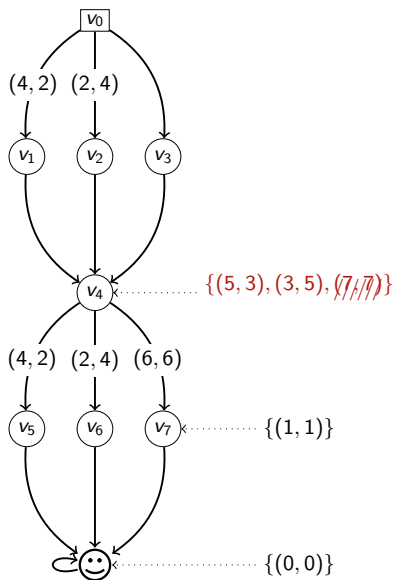




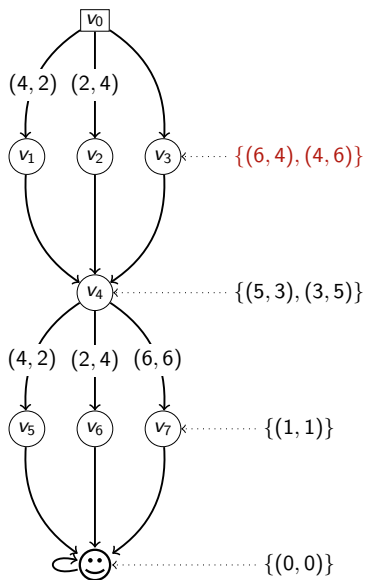
# Computing Pareto( $v$ )



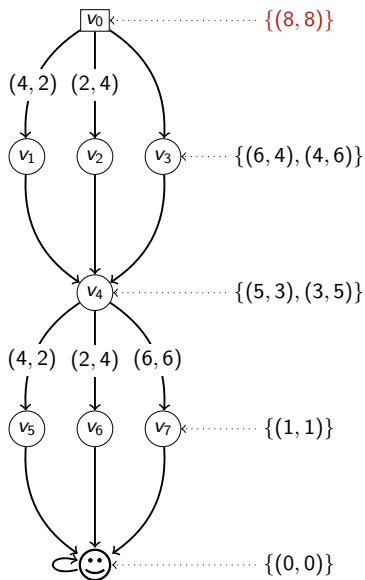
## Computing Pareto( $v$ )



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## Computing Pareto( $v$ )



# Pareto-optimal strategies

for  $v \in F$  do  $I^0(v) = \{0\}$   
 for  $v \notin F$  do  $I^0(v) = \{\infty\}$

repeat

for  $v \in V$  do  
 if  $v \in F$  then  $I^{k+1}(v) = \{0\}$

else if  $v \in V_{\square}$  then

$$I^{k+1}(v) = \text{minimal} \left( \bigcup_{v' \in \text{succ}(v)} \uparrow I^k(v') + \mathbf{w}(v, v') \right)$$

for  $x \in I^{k+1}(v)$  do

if  $x \in I^k(v)$  then  $f_v^{k+1}(x) = f_v^k(x)$

else

$f_v^{k+1}(x) = (v', x')$  where  $v'$  and  $x'$  are such that  $v' \in \text{succ}(v)$ ,  $x = x' + \mathbf{w}(v, v')$  and  $x' \in I^k(v')$

else if  $v \in V_{\square}$  then

$$I^{k+1}(v) = \text{minimal} \left( \bigcap_{v' \in \text{succ}(v)} \uparrow I^k(v') + \mathbf{w}(v, v') \right)$$

until  $I^{k+1}(v) = I^k(v)$  for all  $v \in V$

# Computing Pareto-optimal strategies

Given  $u \in V$  and  $\mathbf{c} \in I^*(u) \setminus \{\infty\}$ , we define a strategy  $\sigma_{\bigcirc}^*$  from  $u$  such that for all  $hv \in \text{Hist}_{\bigcirc}(u)$ , let  $\mathcal{C}(hv) = \{\mathbf{x}' \in I^*(v) \mid \mathbf{x}' \leq_{\mathbf{C}} \mathbf{c} - \text{Cost}(hv) \wedge \mathbf{x}' \leq_{\mathbf{L}} \mathbf{c} - \text{Cost}(hv)\}$ ,

$$\sigma_{\bigcirc}^*(hv) = \begin{cases} v' & \text{for some } v' \in \text{succ}(v), \text{ if } \mathcal{C}(hv) = \emptyset \\ f_v^*(\mathbf{x})[1] & \text{where } \mathbf{x} = \min_{\leq_{\mathbf{L}}} \mathcal{C}(hv), \text{ if } \mathcal{C}(hv) \neq \emptyset \end{cases}.$$

$\sigma_{\bigcirc}^*$  is a  $\mathbf{c}$ -Pareto-optimal strategy from  $u$ .

## Conclusion

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	Componentwise order	Lexicographic order
$\text{minimal}(\text{Ensure}(v))$	in exponential time	in polynomial time
CEP	PSpace-complete	in P

- **uniform approach** to compute  $\text{minimal}(\text{Ensure}(v))$  both for the componentwise order and the lexicographic order  $\rightsquigarrow$  fixpoint algorithm;
- (Pareto)-optimal strategies can be synthesized thanks to the fixpoint algorithm;
- Pareto-optimal strategies may **require memory**.



