

# Mini-course on algorithmic aspects of stochastic games and related models



清华大学  
Tsinghua University

交叉信息研究院  
Institute for Interdisciplinary Information Sciences

CTIC  
交互计算

Marcin Jurdziński (University of Warwick)

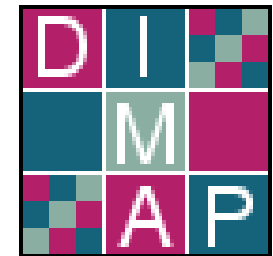
Peter Bro Miltersen (Aarhus University)

Uri Zwick (武熠) (Tel Aviv University)

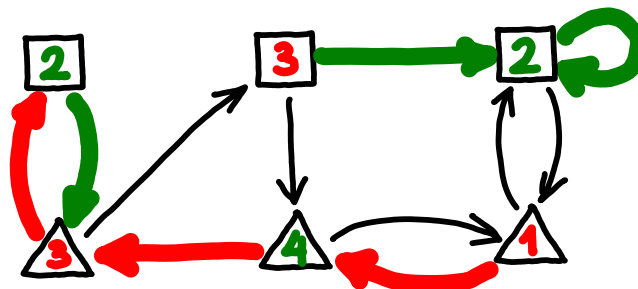
Oct. 31 – Nov. 2, 2011

Day 2  
Tuesday, November 1

Marcin Jurdziński  
(University of Warwick)

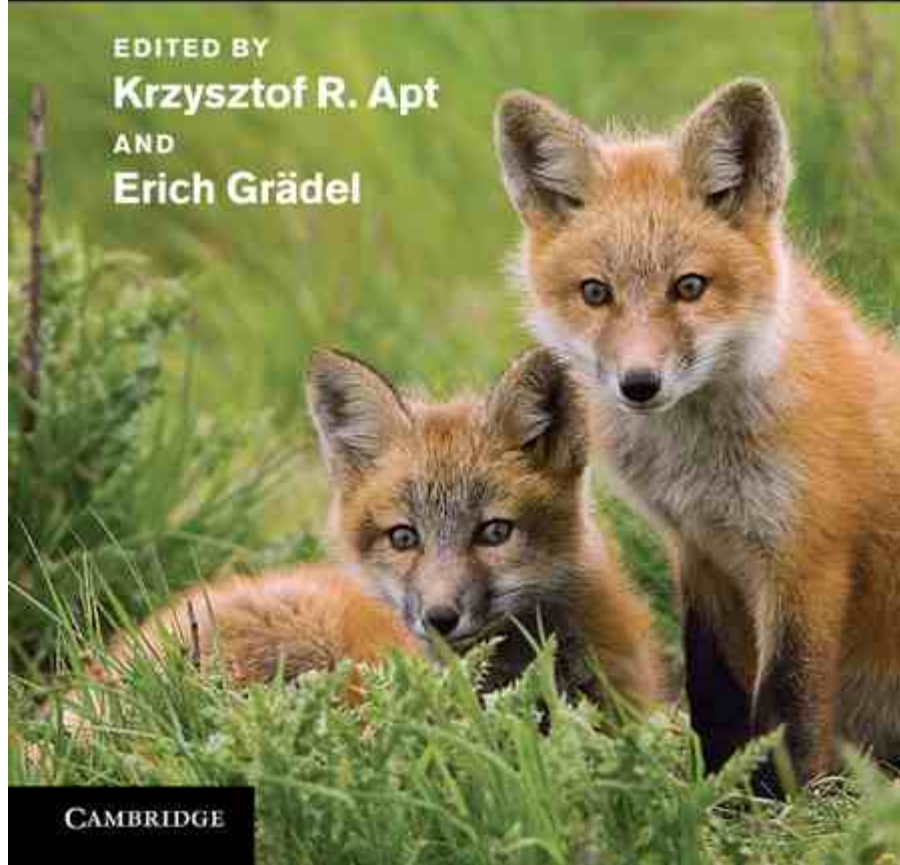


## Parity Games



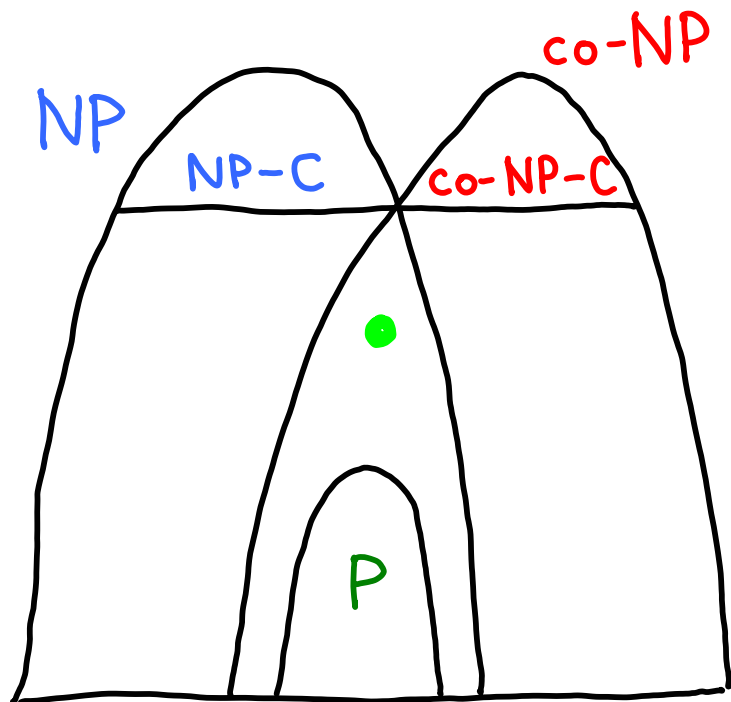
# Lectures in Game Theory for Computer Scientists

EDITED BY  
**Krzysztof R. Apt**  
AND  
**Erich Grädel**

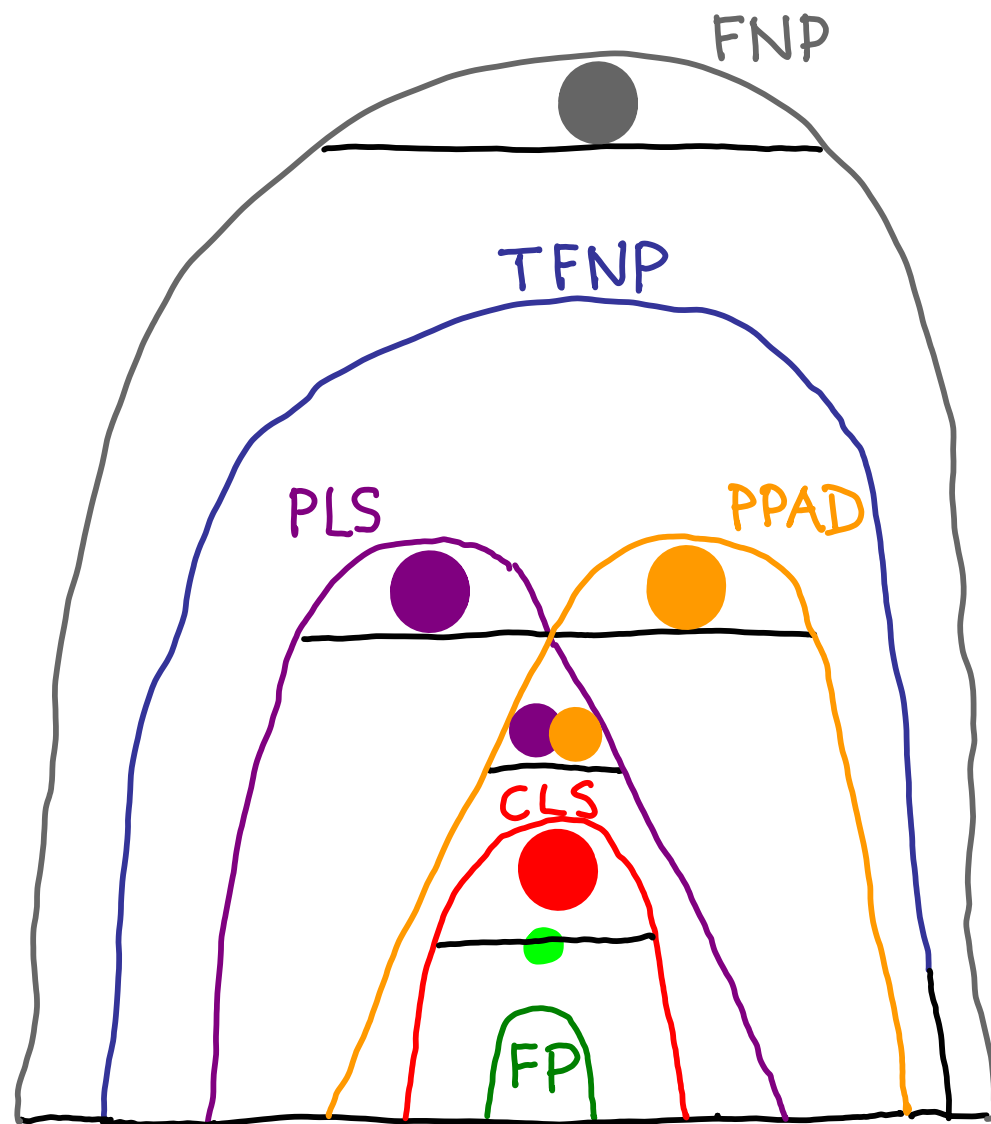


CAMBRIDGE

How DIFFICULT IS IT TO FIND  
A WINNING STRATEGY?



COMPUTING  
OPTIMAL  
STRATEGIES



# PLAN

1. Motivating example
2. Games on graphs
3. Reachability / safety games
4. Büchi /  $\omega$ -Büchi games
5. Parity games
6. Better algorithms for parity games
7. Search complexity of computing optimal strategies

# BOOLEAN FORMULA EVALUATION

State  $s$ :

|   |   |
|---|---|
| P | 0 |
| q | 1 |
| r | 1 |

Property  $\varphi$ :  $(r \vee p) \wedge (\neg p \wedge (q \vee \neg r))$

$$s \stackrel{?}{\models} \varphi$$

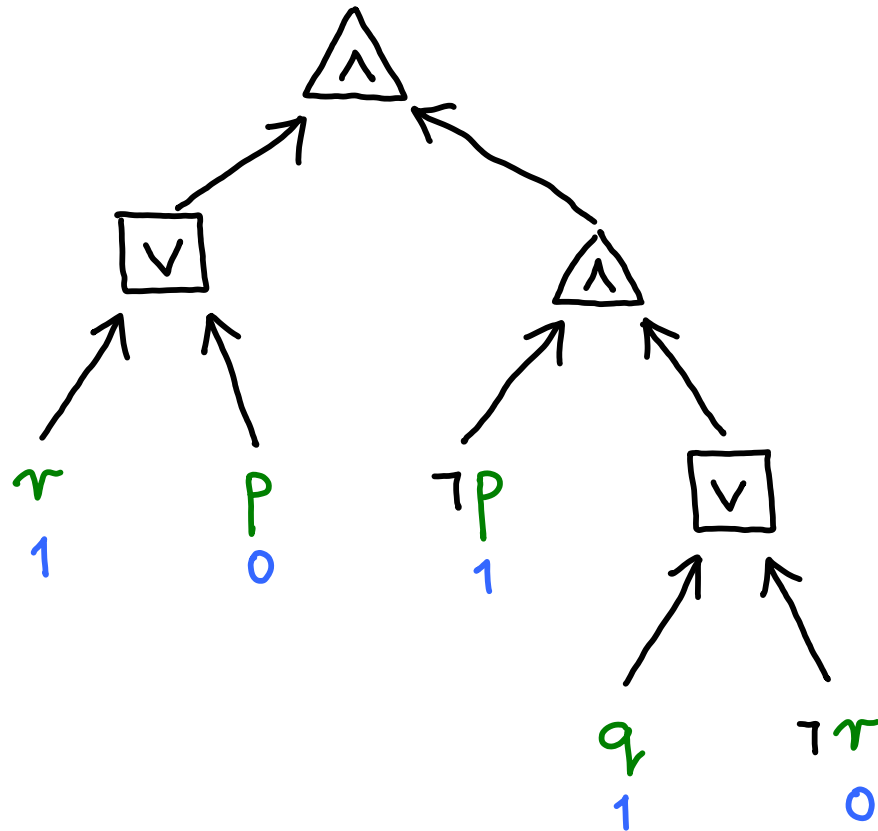
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$C_{s,\varphi}$ :



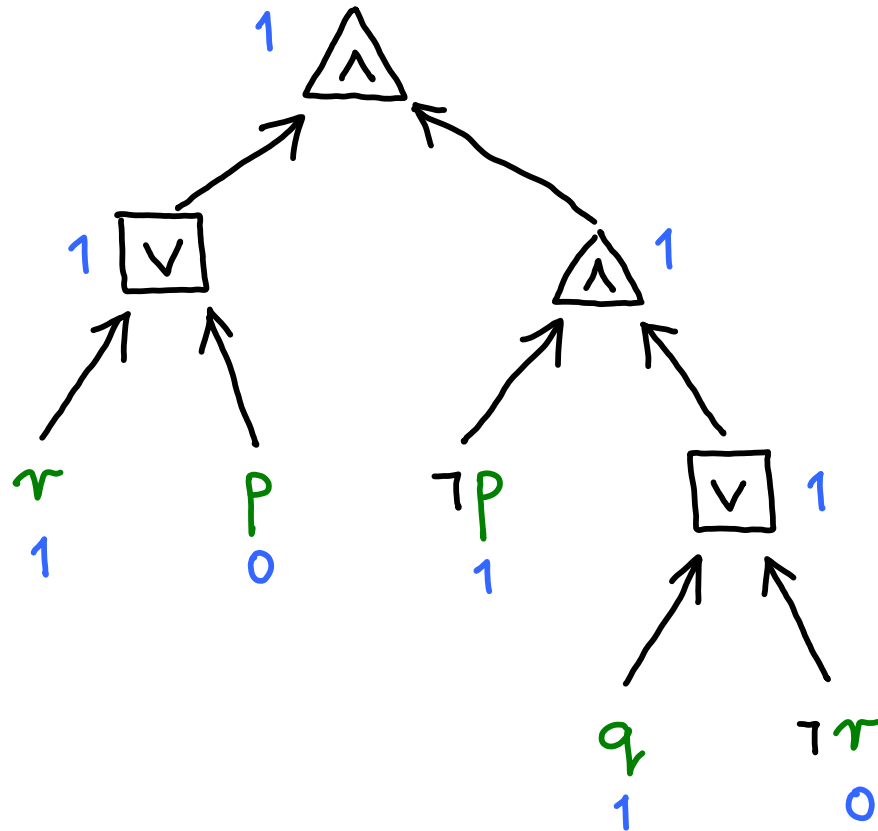
# BOOLEAN FORMULA EVALUATION

State  $s$ :

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|---|---|
| P | 0 |
| q | 1 |
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Property  $\varphi$ :  $(r \vee p) \wedge (\neg p \wedge (q \vee \neg r))$

$\mathcal{C}_{s,\varphi}$ :



**FACT**

$$s \models \varphi$$

iff

the value of  
circuit  $\mathcal{C}_{s,\varphi}$  is 1



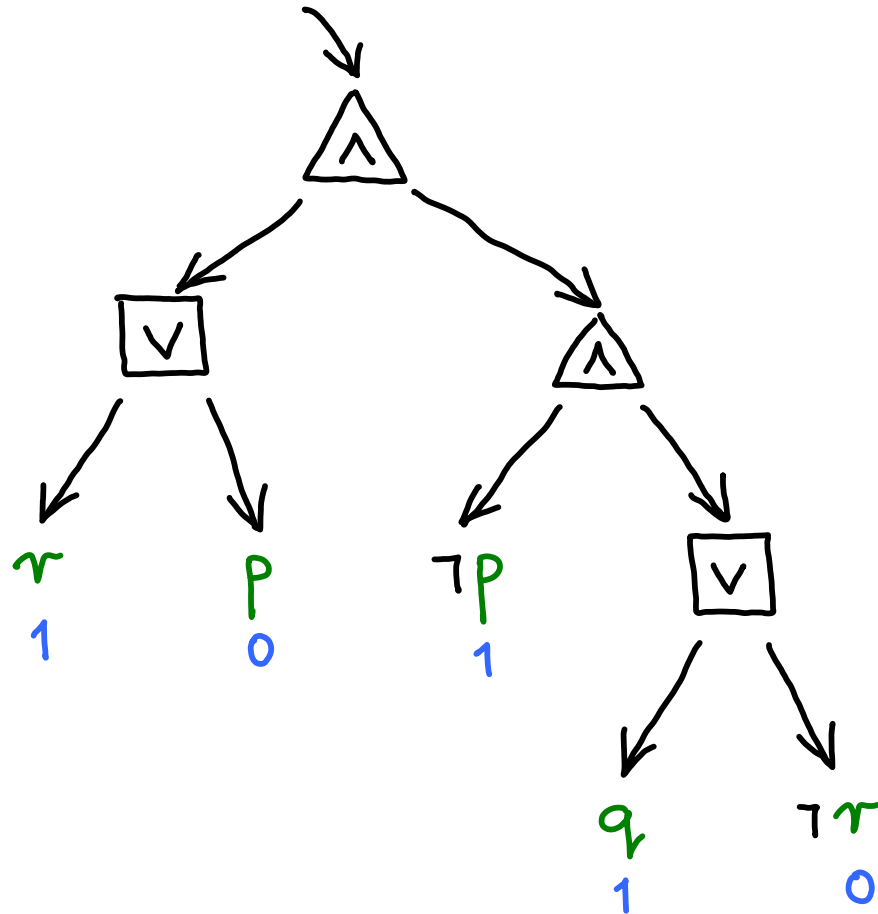
# BOOLEAN FORMULA EVALUATION: GAME

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$G_{s,\varphi}$ :



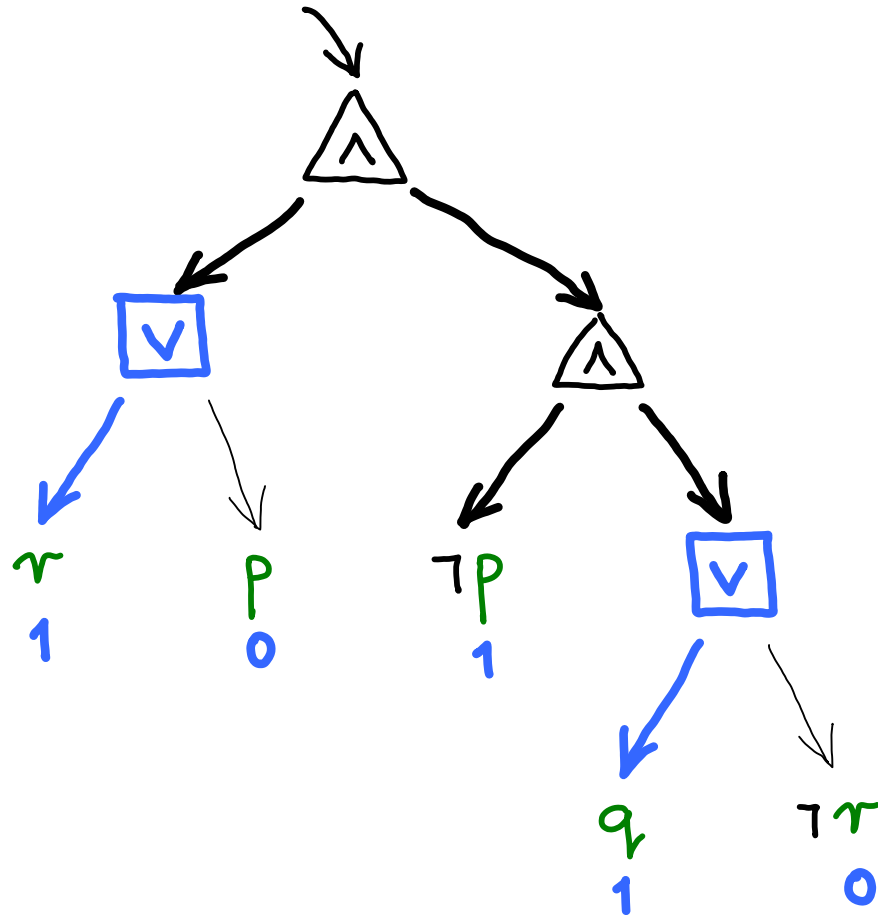
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FACT

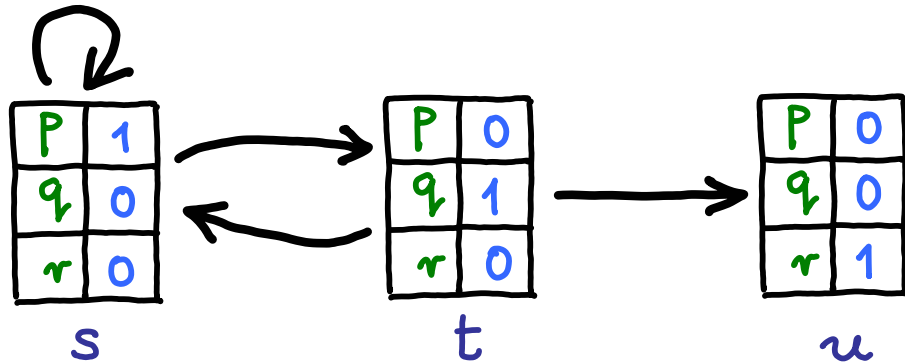
$$s \models \varphi$$

iff

$\square$  has a strategy to reach 1 in  $G_{s,\varphi}$

# TEMPORAL FORMULA EVALUATION

(Model of a) program :



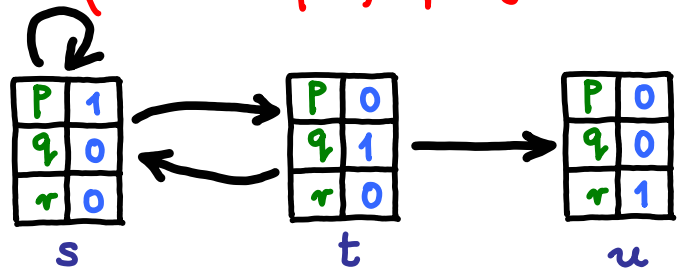
CTL property  $\varphi$ :

$$E((P \vee q) U r)$$

$$s \models ? \varphi$$

# TEMPORAL FORMULA EVALUATION: GAME

(Model of a) program:

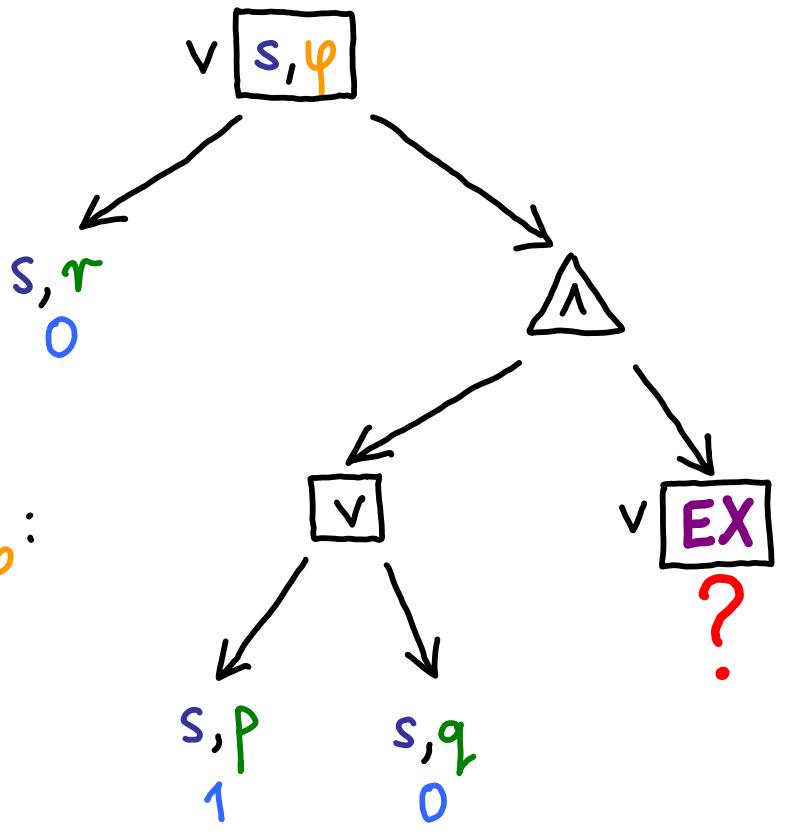


CTL property  $\varphi$ :

$$\varphi: E((p \vee q) U r)$$

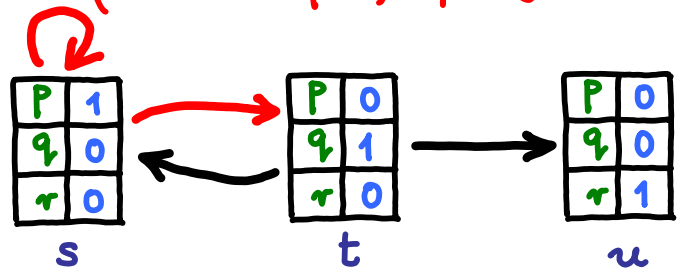
$$\equiv r \vee ((p \vee q) \wedge EX \varphi)$$

$G_{s, \varphi}$ :



# TEMPORAL FORMULA EVALUATION: GAME

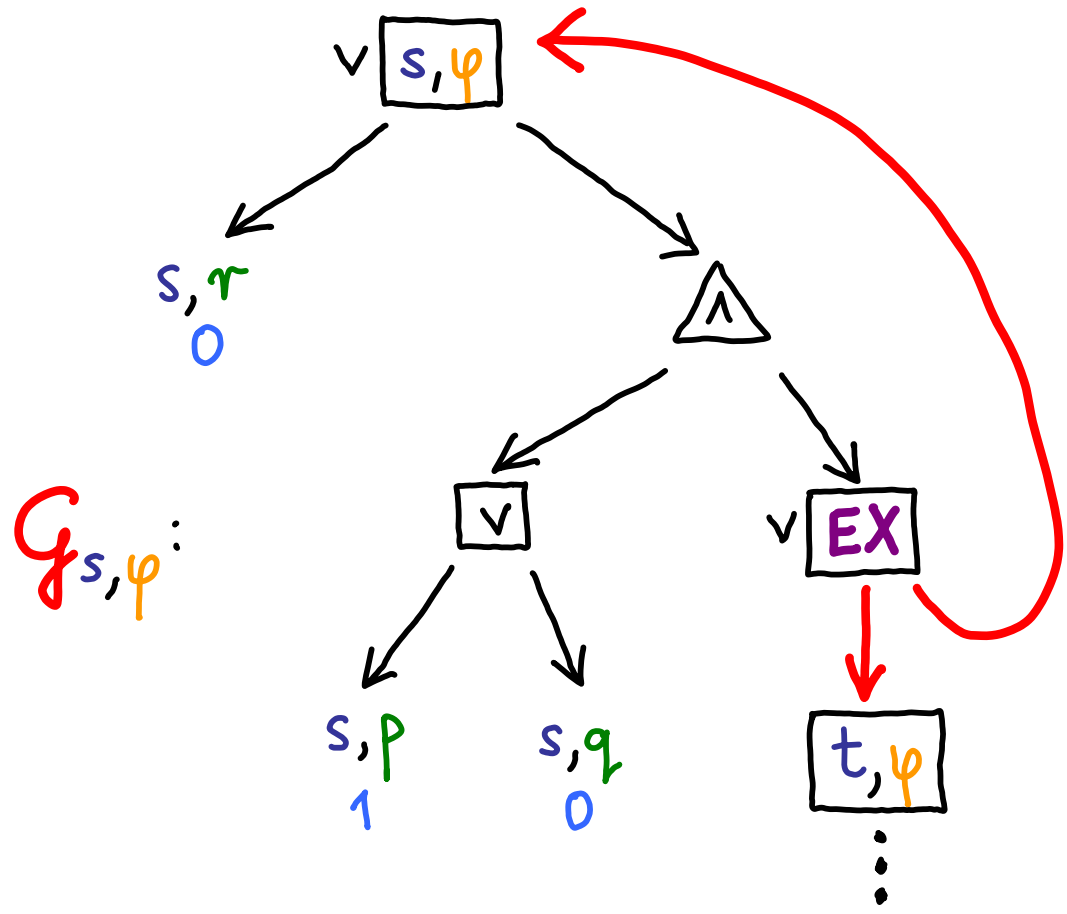
(Model of a) program:



CTL property  $\psi$ :

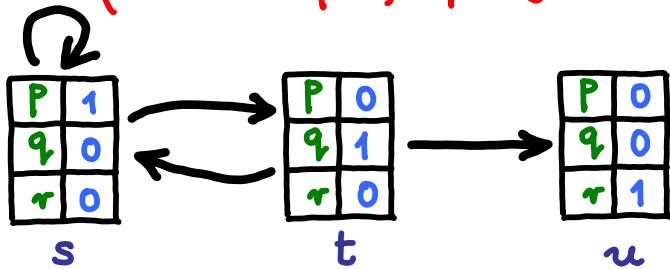
$$\psi: E((p \vee q) U r)$$

$$\equiv r \vee ((p \vee q) \wedge EX \psi)$$



# TEMPORAL FORMULA EVALUATION: GAME

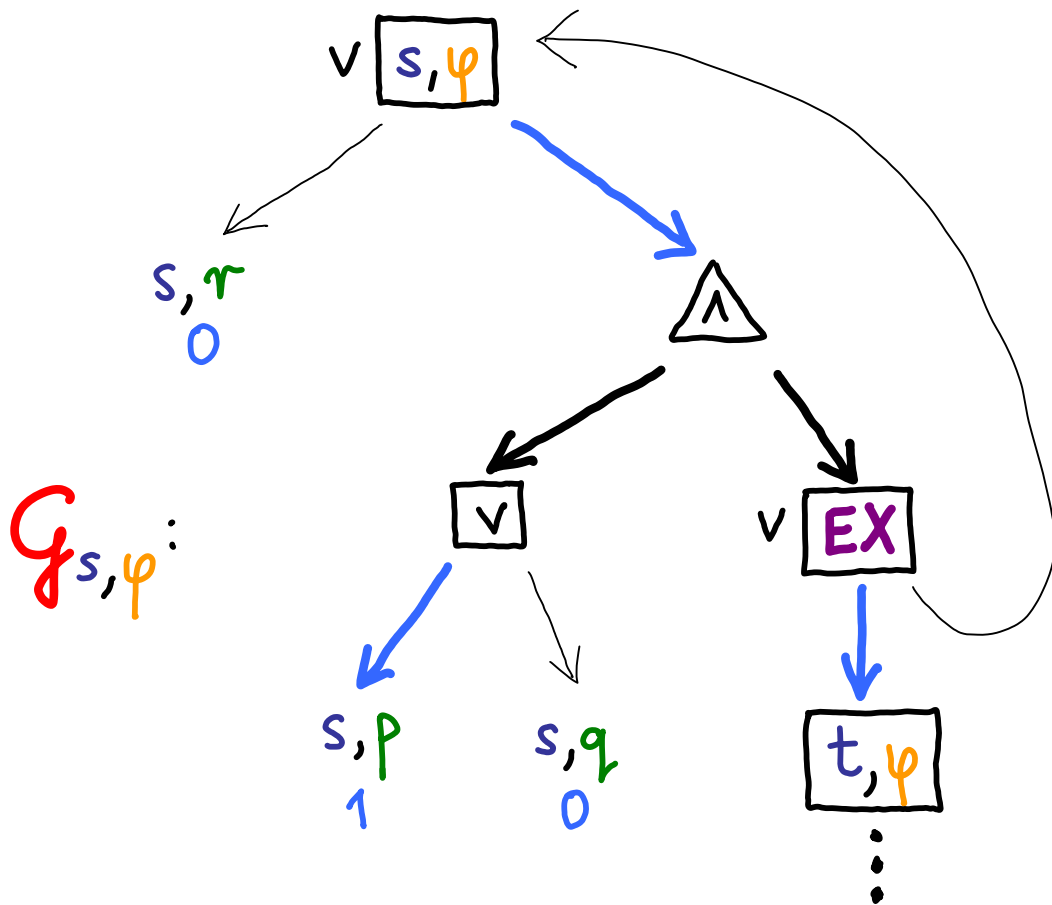
(Model of a) program :



CTL property  $\varphi$ :

$$\varphi: E((p \vee q) U r)$$

$$r \vee ((p \vee q) \wedge EX \varphi)$$



THM

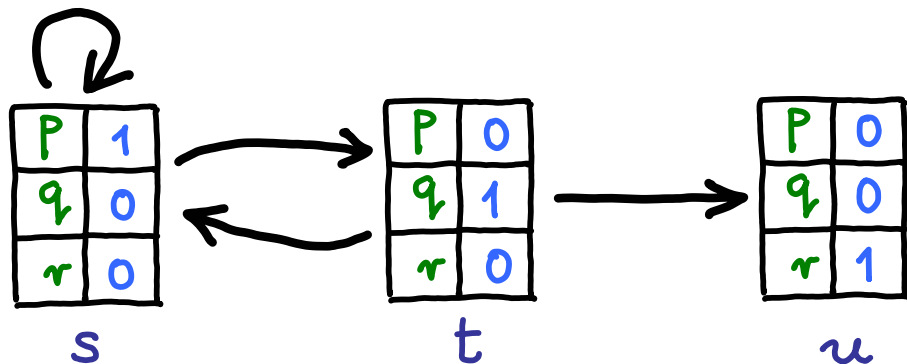
$s \models \varphi$

iff

$\square$  has a strategy to reach 1 in  $G_{s,\varphi}$

# FIXPOINT FORMULA EVALUATION

(Model of a) program :



$\mu$ -calculus property:

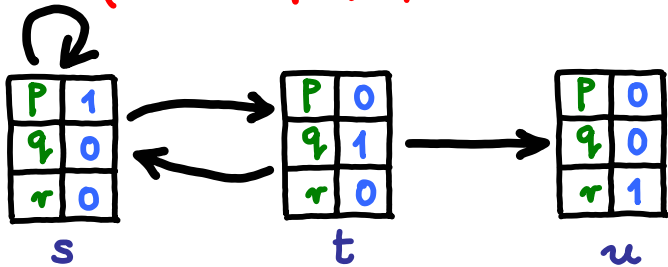
$$\varphi : \boxed{\mu Z. \varphi(Z)}$$

$$\equiv \varphi(\varphi)$$

$$s \stackrel{?}{=} \varphi$$

# FIXPOINT FORMULA EVALUATION

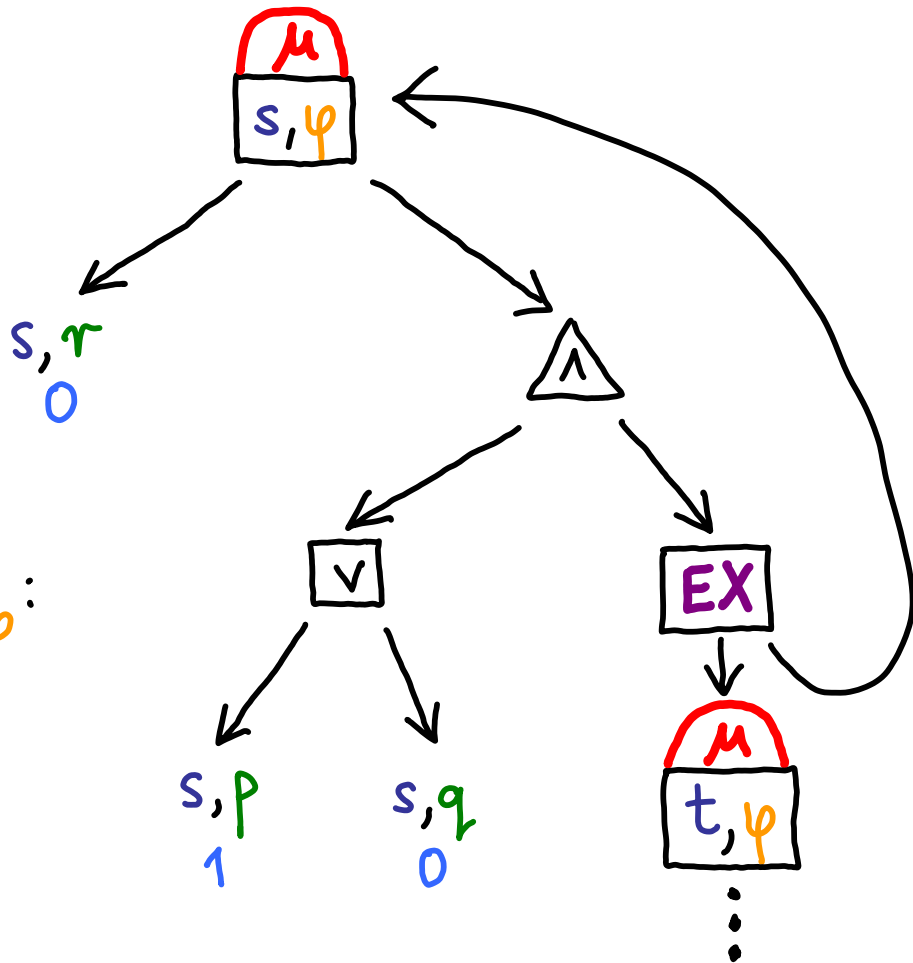
(Model of a) program :



$\mu$ -calculus property  $\varphi$ :

$$\varphi: \boxed{\mu Z. \psi(Z)}$$

$$\equiv \psi(\varphi)$$



## THM

$$s \models \varphi$$

iff

$\square \forall$  has a strategy in  $G_{s, \varphi}$

- to reach 1, or
- outermost infinitely occurring fixpoint subformula is not a  $\mu$  (i.e., least fixpoint)

$G_{s, \varphi}$ :

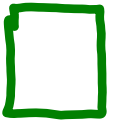


# PLAN

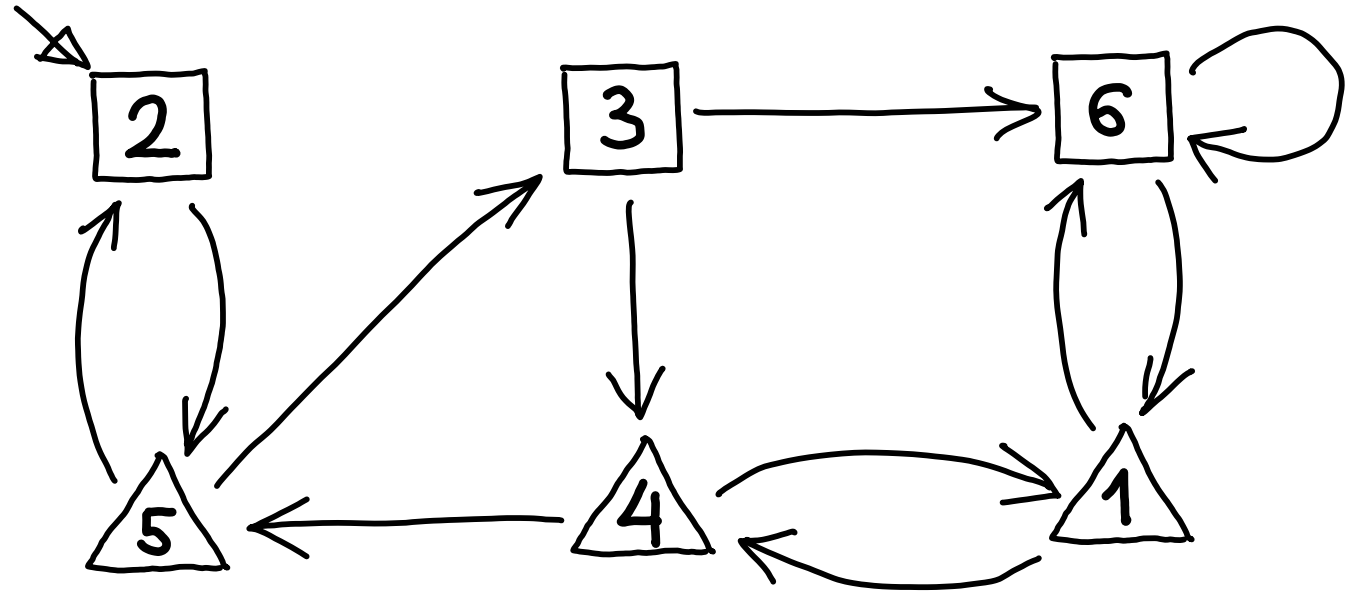
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# GAMES ON GRAPHS: GAME GRAPH

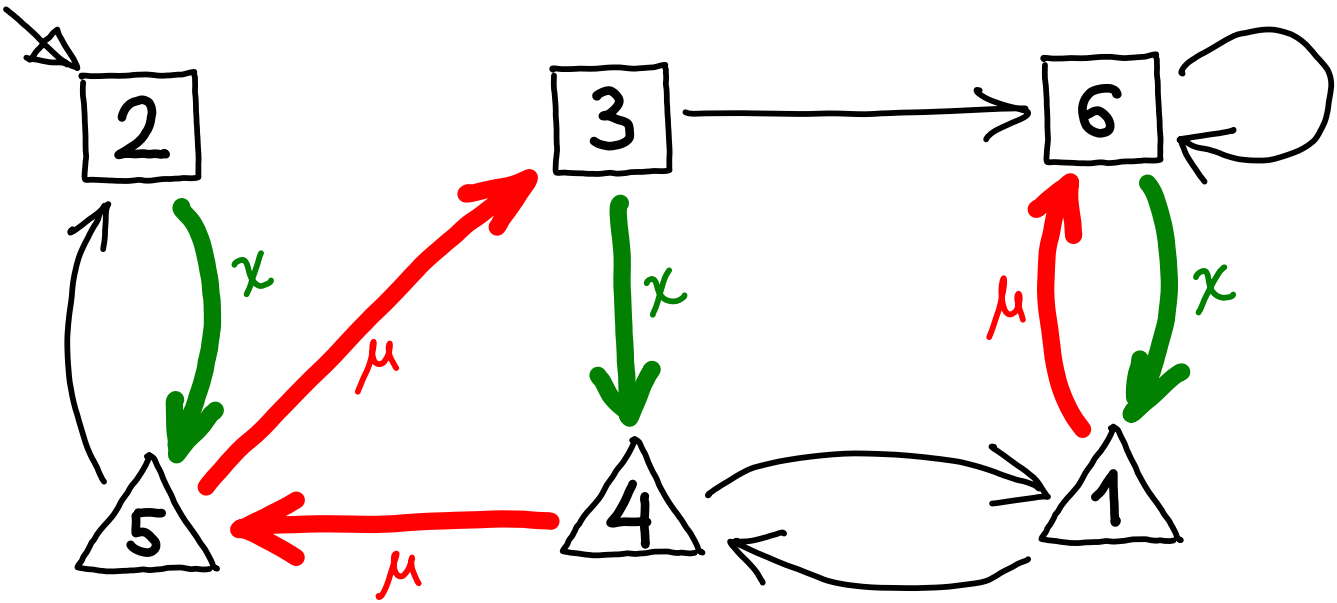
Players:



$n = |V|$  vertices,  $m = |E|$  edges



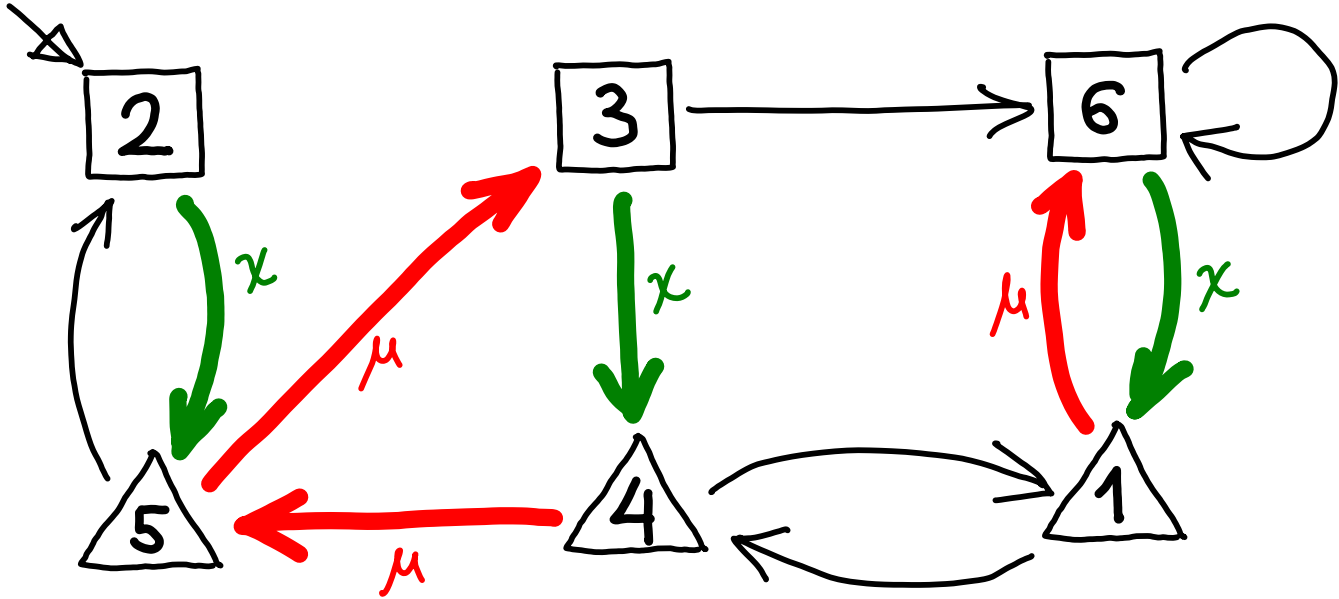
# GAMES ON GRAPHS: POSITIONAL STRATEGIES



Positional strategy for  $\square$ :  $\chi: V_{\square} \rightarrow V$

Positional strategy for  $\triangle$ :  $\mu: V_{\triangle} \rightarrow V$

# GAMES ON GRAPHS: PLAYS



$$\text{Play}(2, x, \mu) = (2, 5, 3, 4, 5, (3, 4, 5)^\omega) \in V^\omega$$

# GAMES ON GRAPHS: OBJECTIVES

Objectives:  $\mathcal{W}_\square \subseteq V^\omega$   
 $\mathcal{W}_\Delta = V^\omega \setminus \mathcal{W}_\square$

# GAMES ON GRAPHS: OBJECTIVES

Objectives:  $W_{\square} \subseteq V^{\omega}$   
 $W_{\Delta} = V^{\omega} \setminus W_{\square}$

Examples of objectives:

- **Reachability:**  $W_{\square} = \{(v_0, v_1, v_2, \dots) : v_i \in T \text{ for some } i \in \mathbb{N}\}$
- **Safety:**  $W_{\square} = \{(v_0, v_1, v_2, \dots) : v_i \in S \text{ for all } i \in \mathbb{N}\}$

where  $T, S \subseteq V$

# GAMES ON GRAPHS: WINNING STRATEGIES

$\chi \in \Sigma_{\square}$  is a *winning strategy* for  $\square$  from  $v \in V$ ,  
if  $\text{Play}(v, \chi, \mu) \in W_{\square}$ , for all  $\mu \in \Sigma_{\Delta}$

$\mu \in \Sigma_{\Delta}$  is a *winning strategy* for  $\Delta$  from  $v \in V$ ,  
if  $\text{Play}(v, \chi, \mu) \in W_{\Delta}$ , for all  $\chi \in \Sigma_{\square}$

# GAMES ON GRAPHS: ALGORITHMIC PROBLEM

Given: game graph  $(V = V_{\square} \uplus V_{\triangle}, E)$

starting vertex  $v \in V$

objective  $W_{\square} \subseteq V^{\omega}$

answer: if  $\square$  has a winning strategy from  $v$



# GAMES ON GRAPHS: DETERMINACY

A game  $(V = V_{\square} \uplus V_{\triangle}, E; \mathcal{W}_{\square})$  is *determined*

if for every starting vertex  $v \in V$

- $\square$  has a *winning strategy* from  $v$ , *or*
- $\triangle$  has a *winning strategy* from  $v$

# MARTIN'S DETERMINACY THEOREM

THM [MARTIN 1975]

Every game  $(V = V_{\square} \uplus V_{\triangle}, E; W_{\square})$ ,  
such that  $W_{\square} \subseteq V^{\omega}$  is a Borel set,  
is determined

# GAMES ON GRAPHS: ALGORITHMIC PROBLEM

Given: game graph  $(V = V_{\square} \uplus V_{\Delta}, E)$   
objective  $W_{\square} \subseteq V^{\omega}$

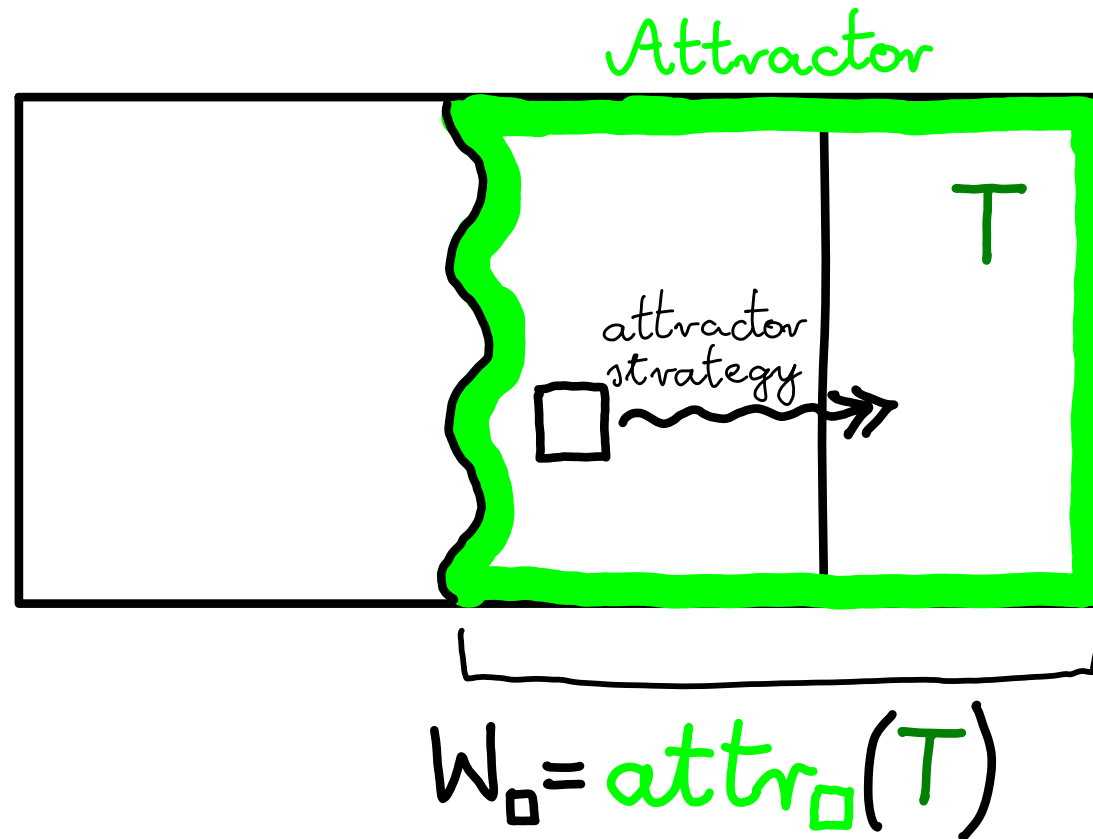
compute:  $V = W_{\square} \uplus W_{\Delta}$  such that

- $\square$  has a winning strategy from  $W_{\square}$ , and
- $\Delta$  has a winning strategy from  $W_{\Delta}$

# PLAN

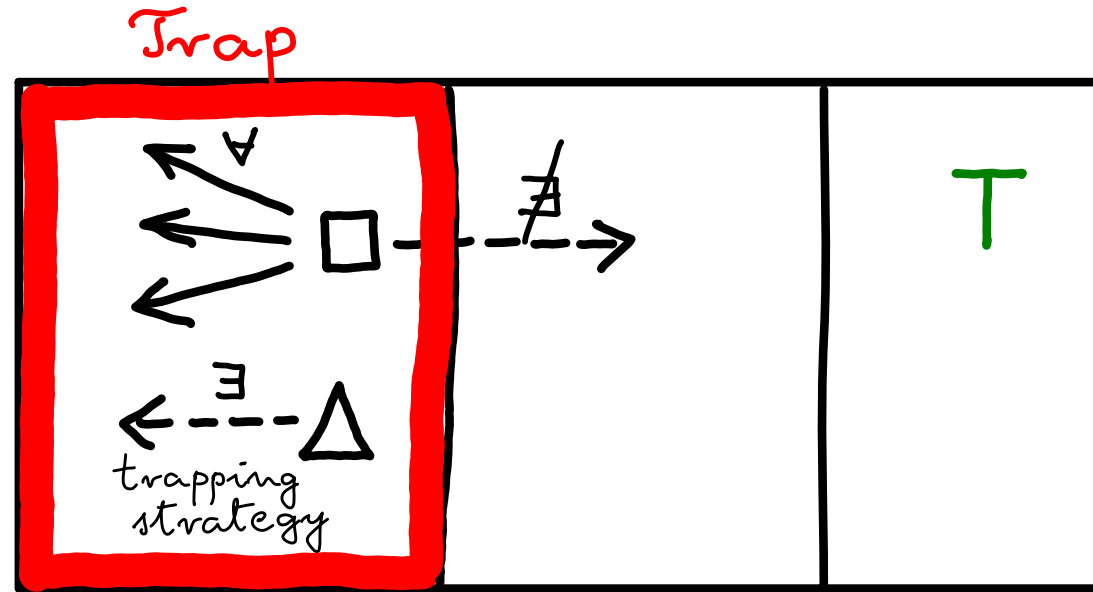
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# REACHABILITY GAMES



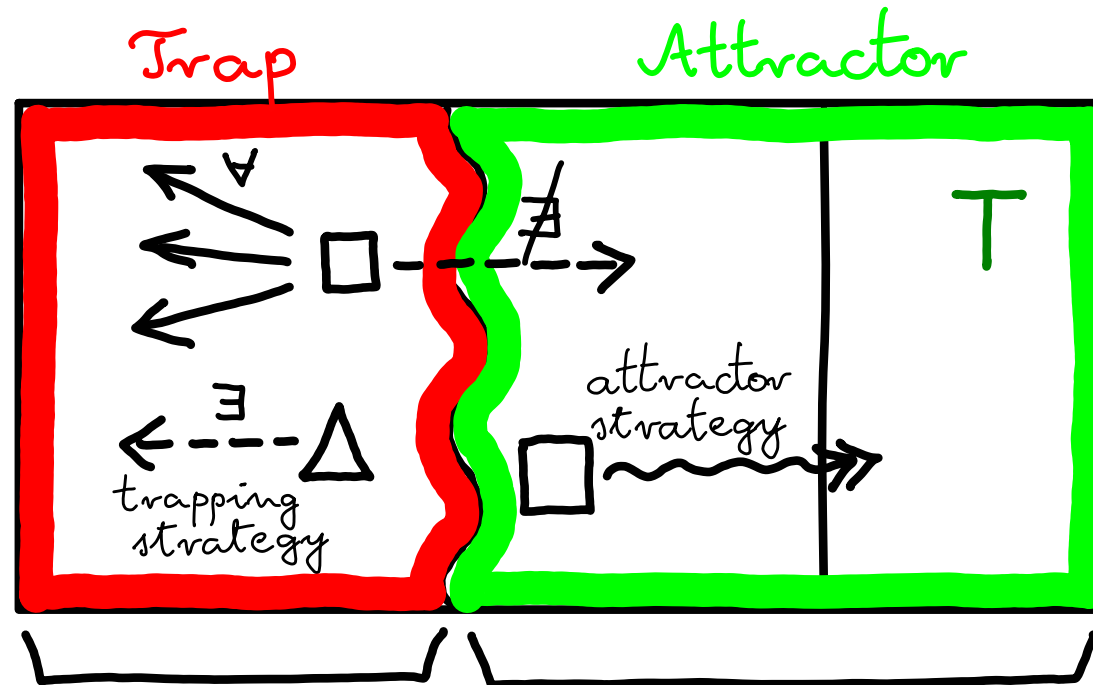
- **Reachability:**  $W_{\square} = \{(v_0, v_1, v_2, \dots) : v_i \in T \text{ for some } i \in \mathbb{N}\}$

# SAFETY GAMES



- **Reachability:**  $W_{\square} = \{(v_0, v_1, v_2, \dots) : v_i \in T \text{ for some } i \in \mathbb{N}\}$
- **Safety:**  $W_{\Delta} = \{(v_0, v_1, v_2, \dots) : v_i \in T \text{ for no } i \in \mathbb{N}\}$

# REACHABILITY/SAFETY GAMES



$$W_{\Delta} = V \setminus \text{attr}_{\square}(T) \quad W_{\square} = \text{attr}_{\square}(T)$$

FACT

Reachability/safety games are positionally determined

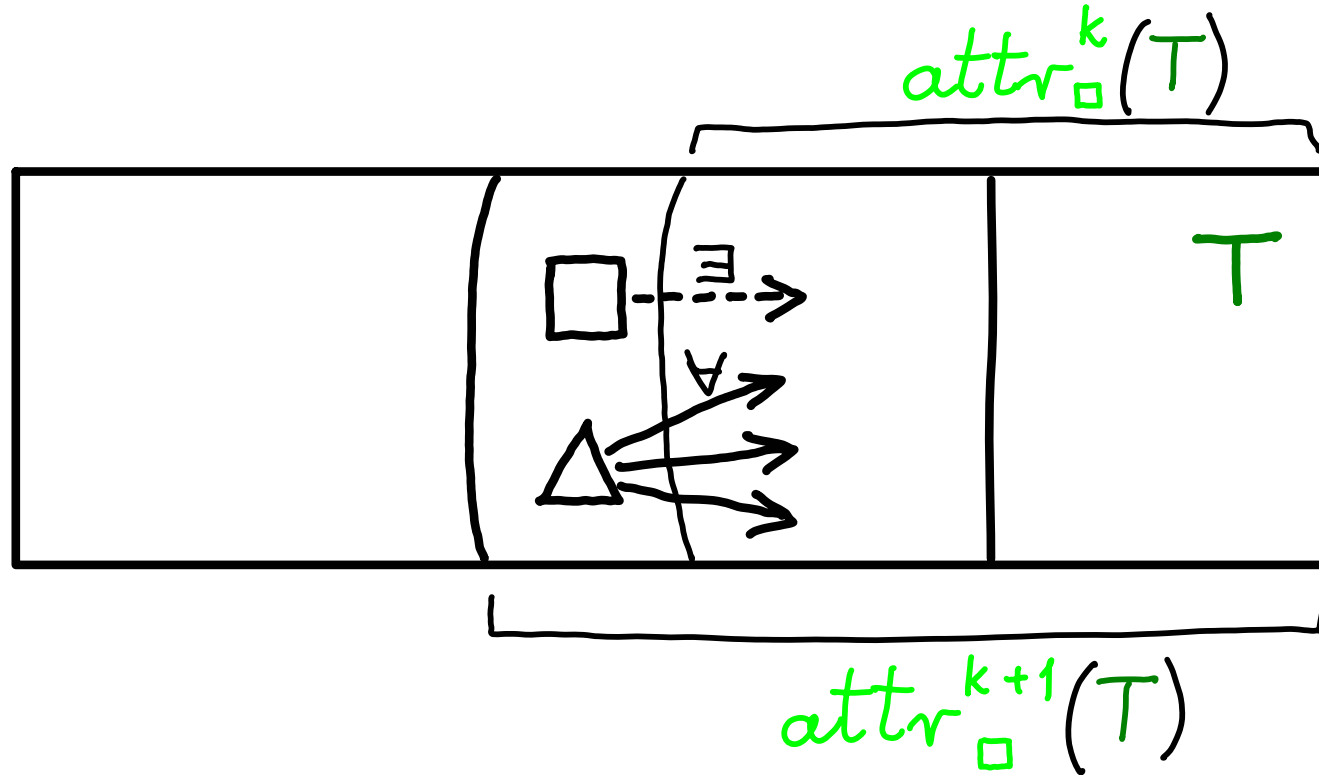
# EXERCISE 1

Explain in detail why the complement of an attractor is a trap for the other player, and how to compute positional attractor and trapping strategies.

Conclude that reachability/safety games are positionally determined.



# SOLVING REACHABILITY/SAFETY GAMES: COMPUTING ATTRACTORS



FACT

$$attr_{\square}(T) = \bigcup_{k \geq 0} attr_{\square}^k(T)$$

can be computed in  $O(m)$  time

## EXERCISE 2

Discuss implementation details that make it possible to compute attractors in time  $O(m)$ .

# PLAN

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# GAMES ON GRAPHS: (co-)BÜCHI OBJECTIVES

- Büchi (repeated reachability)  
An infinite play is winning if  
vertices in  $T \subseteq V$  occur infinitely often
- co-Büchi (eventual safety)  
An infinite play is winning if  
vertices in  $T \subseteq V$  occur finitely often

# ORACLE LEMMA

G



LEMMA

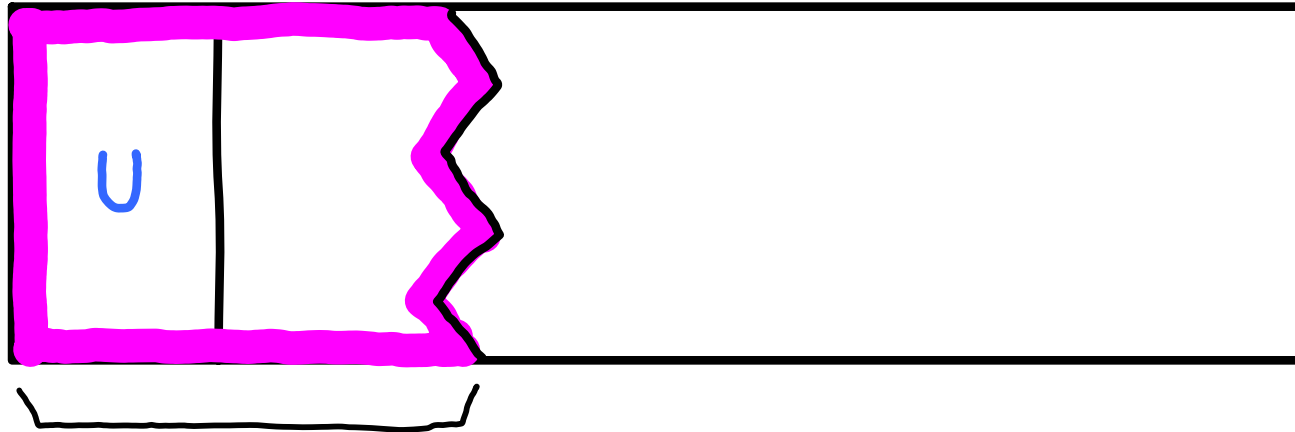
If  $U \subseteq \text{win}_{\Delta}(G)$

then  $\text{win}_{\Delta}(G) = U^* \cup \text{win}_{\Delta}(G'')$

$\text{win}_{\square}(G) = \text{win}_{\square}(G'')$

# ORACLE LEMMA

G



$$U^* = \text{attr}_\Delta(U)$$

## LEMMA

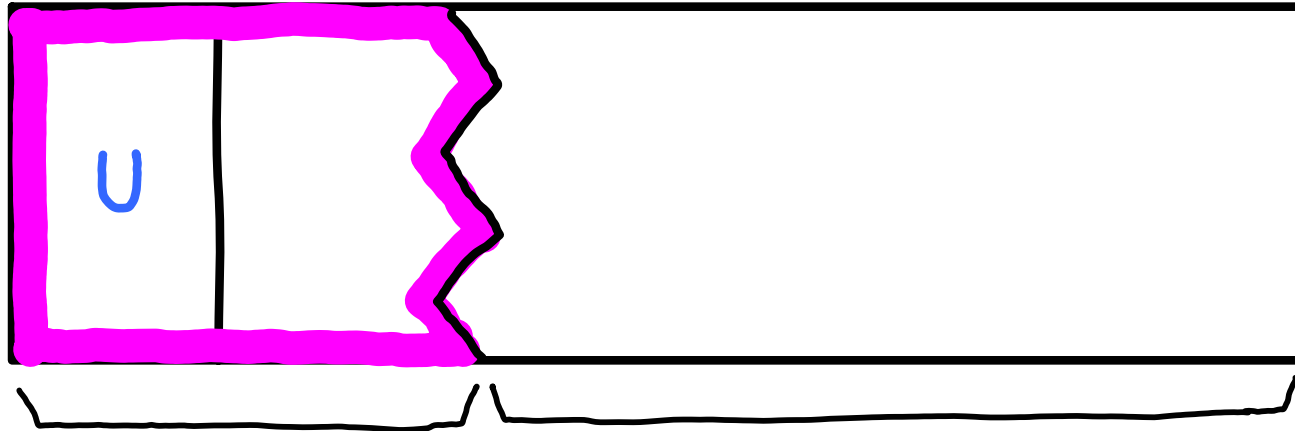
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# ORACLE LEMMA

G



$$U^* = \text{attr}_\Delta(U)$$

$$G'' = G \setminus U^*$$

LEMMA

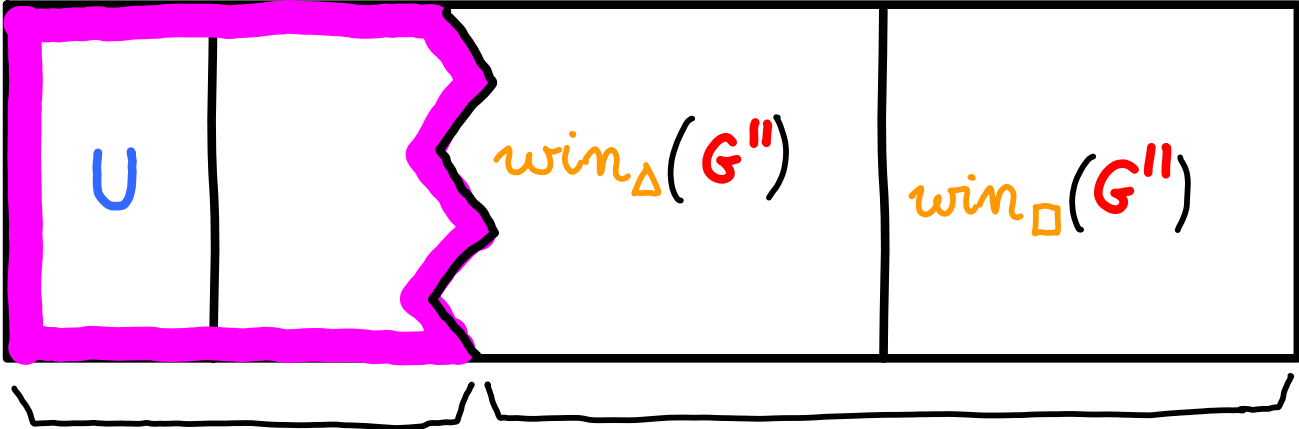
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# ORACLE LEMMA

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$$U^* = \text{attr}_\Delta(U)$$

$$G'' = G \setminus U^*$$

LEMMA

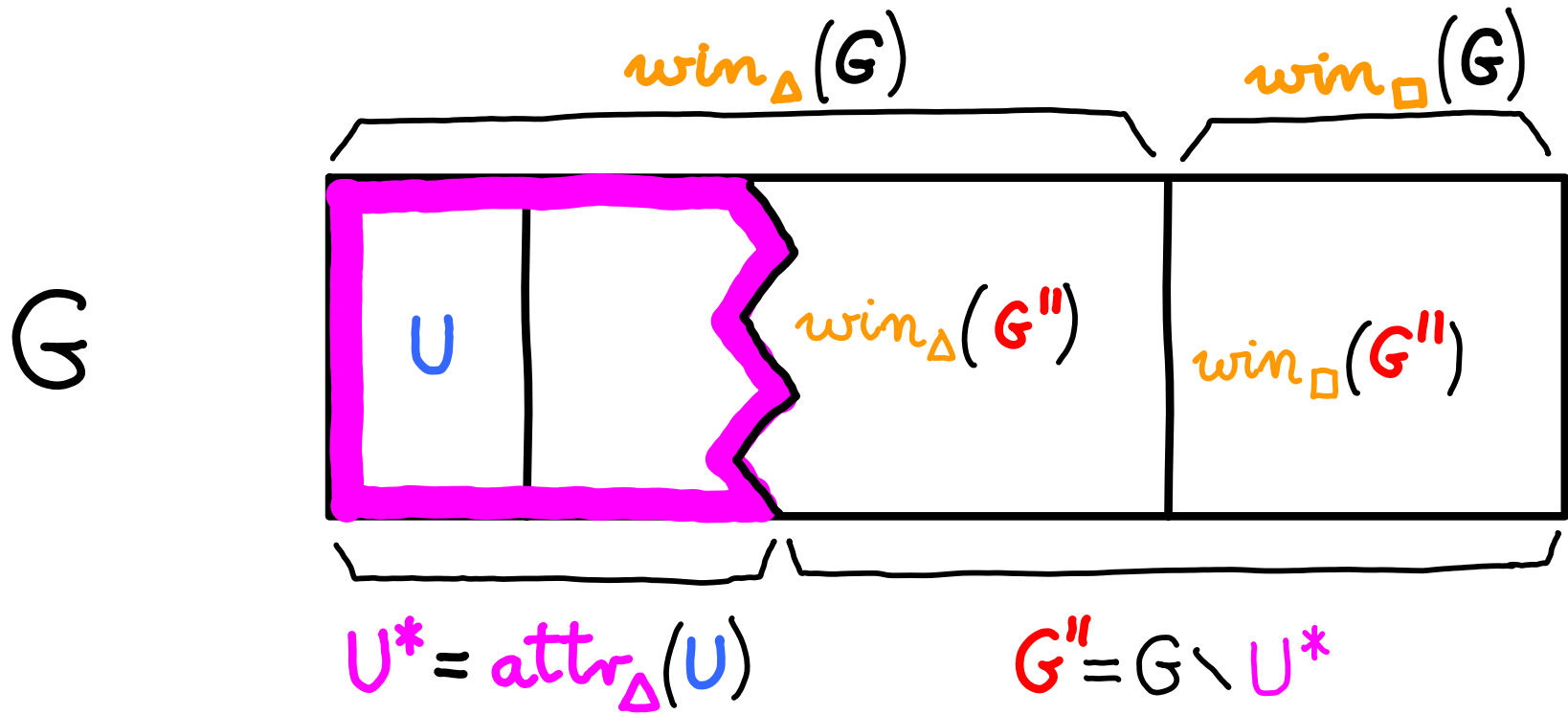
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# ORACLE LEMMA



## LEMMA

If  $U \subseteq win_{\Delta}(G)$

then  $win_{\Delta}(G) = U^* \cup win_{\Delta}(G'')$

$win_{\square}(G) = win_{\square}(G'')$

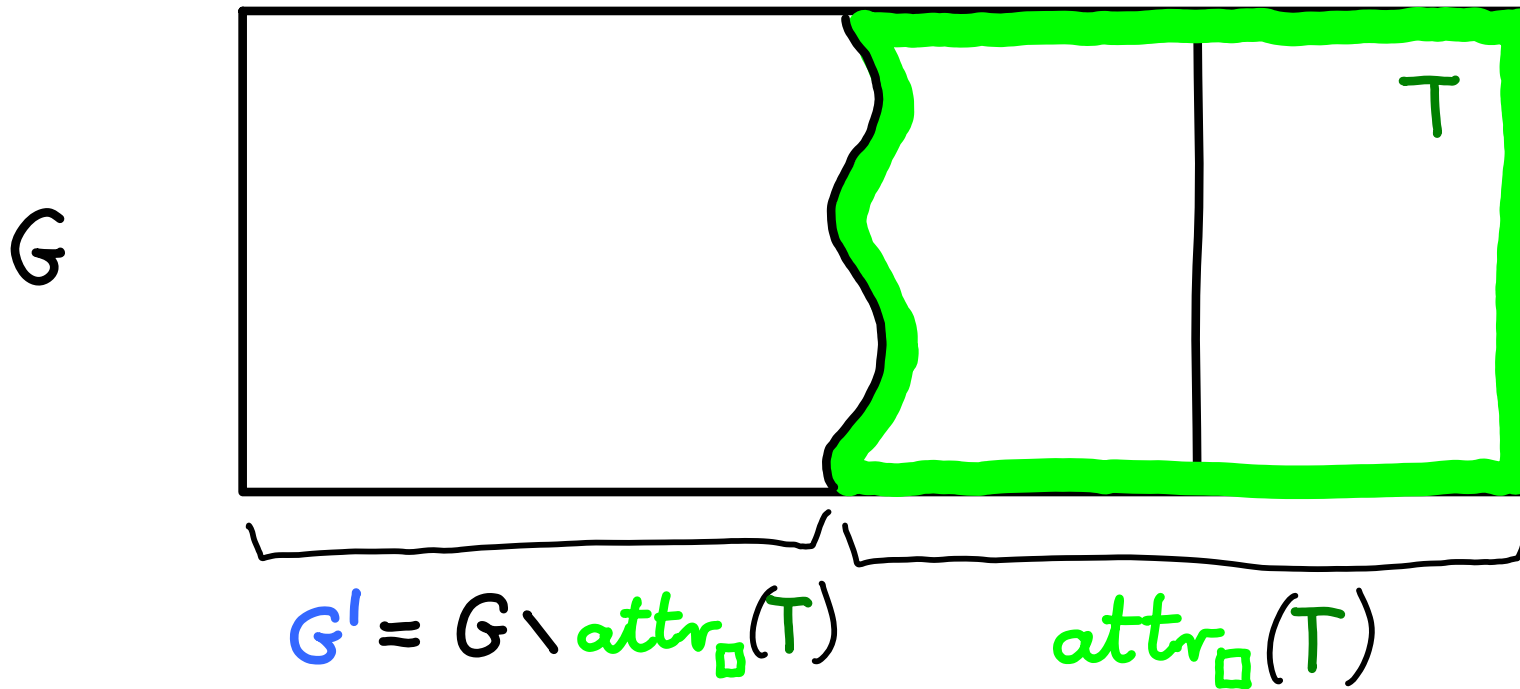
## EXERCISE 3

Prove the oracle lemma  
for Büchi and co-Büchi games.

Are the winning strategies constructed positional?

If not, what assumptions do you need to make  
them so?

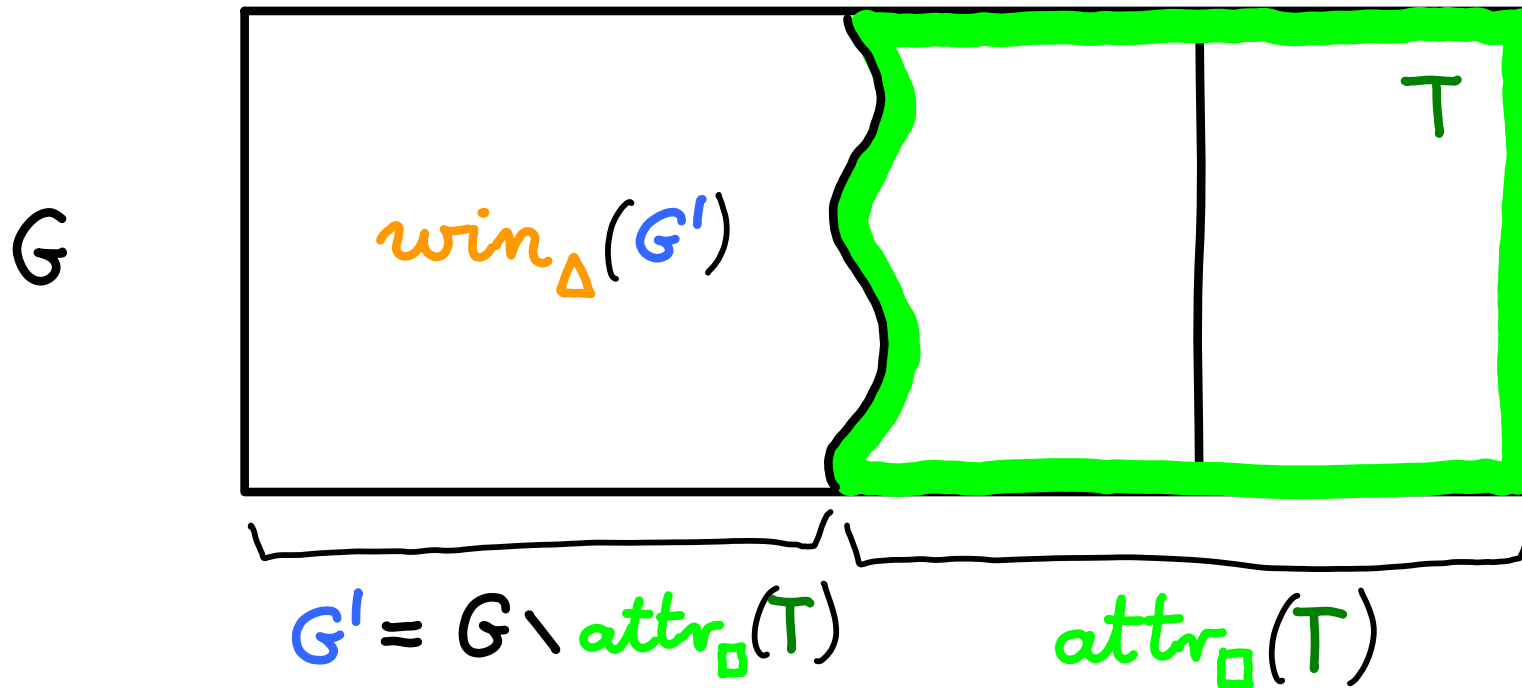
# T-ATTRACTOR LEMMA



## LEMMA

1.  $\text{win}_\Delta(G') = V \setminus \text{attr}_\square(T) \subseteq \text{win}_\Delta(G)$
2. If  $G' = \emptyset$  then  $\text{win}_\square(G) = V$

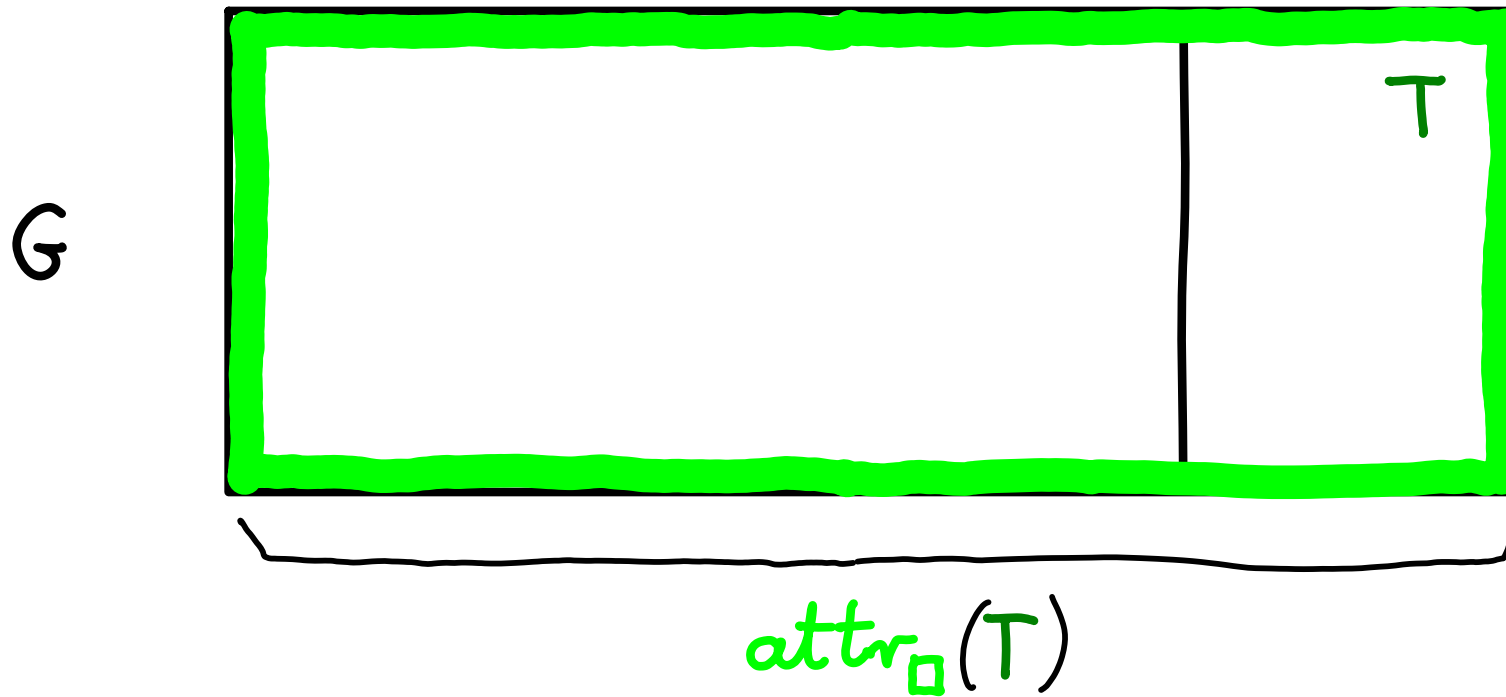
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# T-ATTRACTOR LEMMA



## LEMMA

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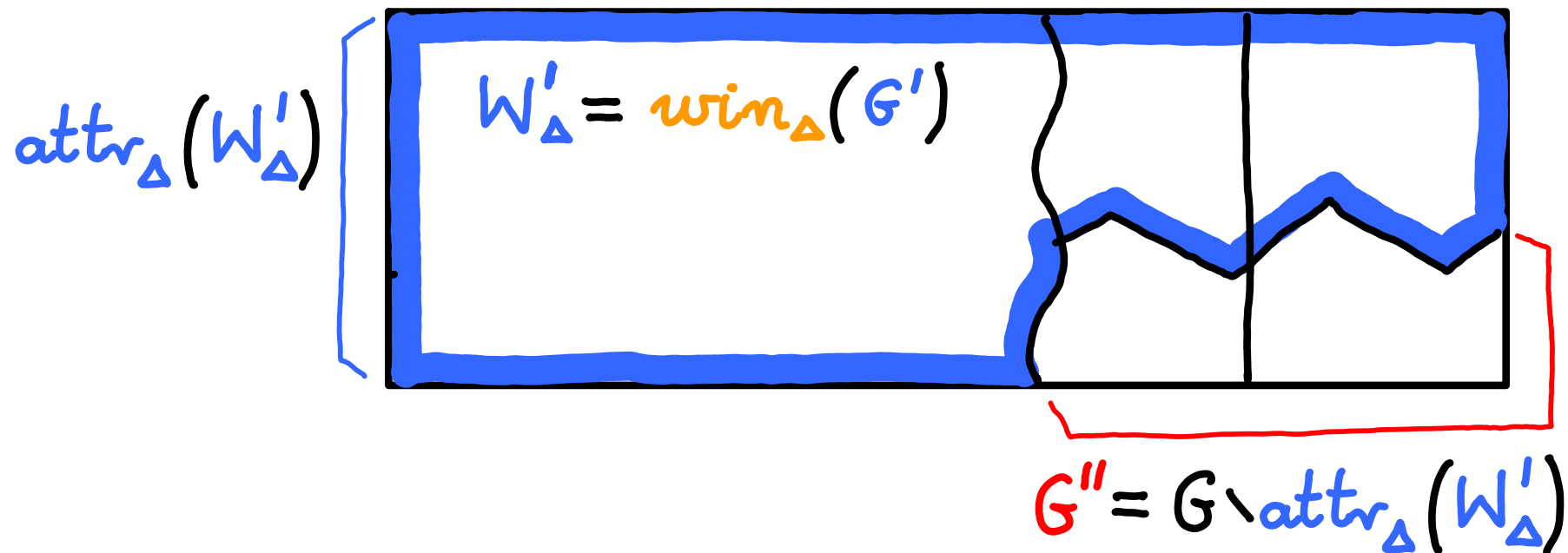
2. If  $G' = \emptyset$  then  $\text{win}_{\square}(G) = V$

# A DIVIDE-AND-CONQUER ALGORITHM



WIN(G): if  $G' = \emptyset$  then  $(W_{\square}, W_{\Delta}) := (V, \emptyset)$   
else  $(W'_{\square}, W'_{\Delta}) := WIN(G'')$   
 $(W_{\square}, W_{\Delta}) := (W'_{\square}, attr_{\Delta}(W'_{\Delta}) \cup W'_{\Delta})$   
return  $(W_{\square}, W_{\Delta})$

# A DIVIDE-AND-CONQUER ALGORITHM



WIN(G): if  $G' = \emptyset$  then  $(W_\square, W_\Delta) := (v, \emptyset)$   
else  $(W''_\square, W''_\Delta) := WIN(G'')$   
 $(W_\square, W_\Delta) := (W''_\square, attr_\Delta(W'_\Delta) \cup W''_\Delta)$   
return  $(W_\square, W_\Delta)$

# POSITIONAL DETERMINACY OF (co-)BÜCHI GAMES

## LEMMA

Büchi/co-Büchi games are positionally determined  
(both on finite and infinite graphs)



## EXERCISE 4

Prove the positional determinacy of Büchi and co-Büchi games.

# RUNNING TIME OF THE DIVIDE-AND-CONQUER ALGORITHM

$T(n)$

WIN(G):

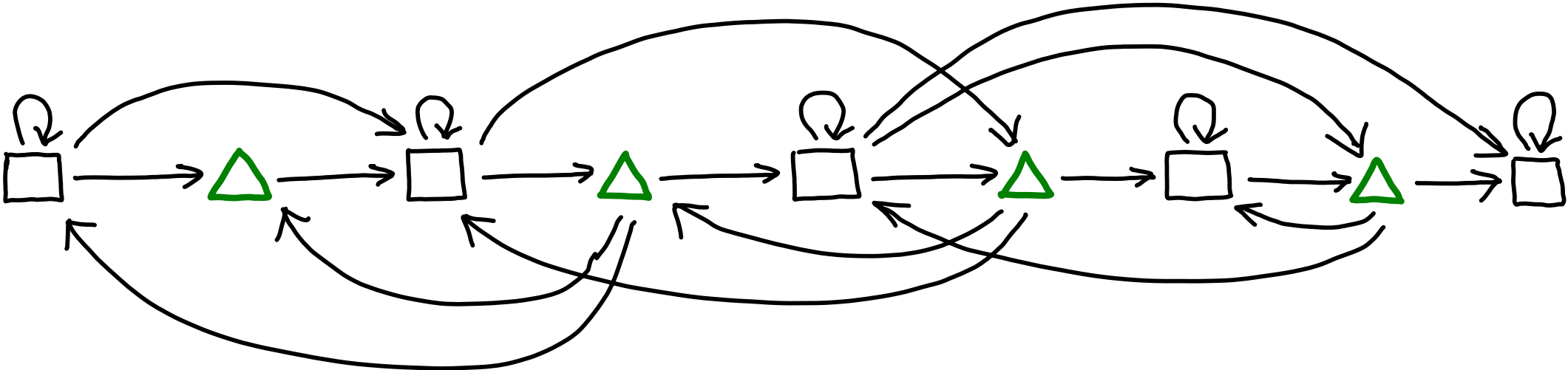
if  $G' = \emptyset$  then  $(W_{\square}, W_{\Delta}) := (V, \emptyset)$   
    else  $(W_{\square}'' , W_{\Delta}'' ) := \underline{\text{WIN}(G'')}$   
     $(W_{\square}, W_{\Delta}) := (W_{\square}'' , \text{attr}_{\Delta}(W_{\Delta}') \cup W_{\Delta}'')$   
return  $(W_{\square}, W_{\Delta})$

$T(n-1)$

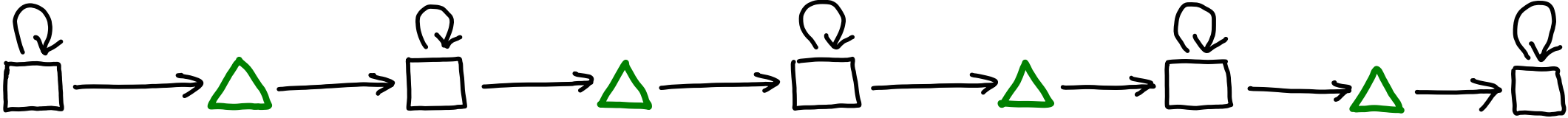
Recurrence:  $T(n) \leq T(n-1) + O(m)$

Solution:  $T(n) = O(n \cdot m)$

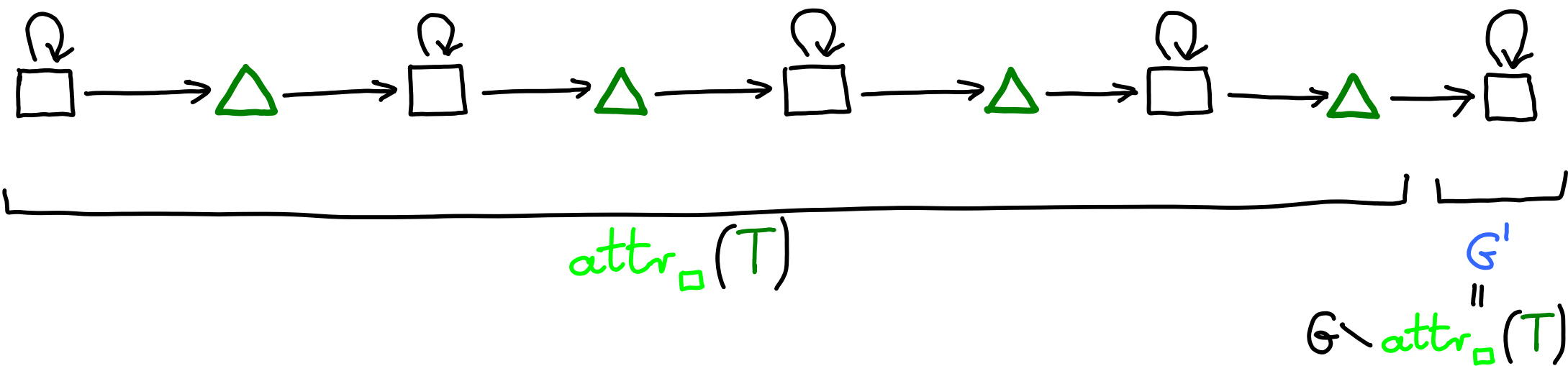
A BÜCHI GAME



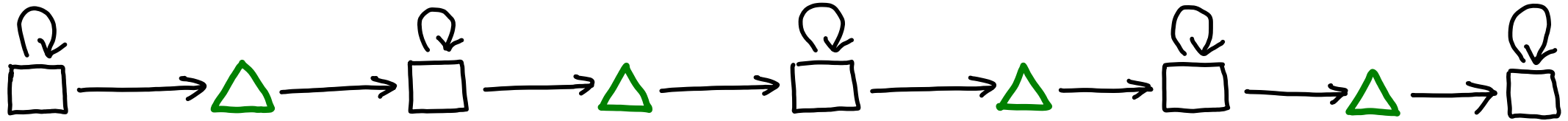
# A BÜCHI GAME



# A BÜCHI GAME



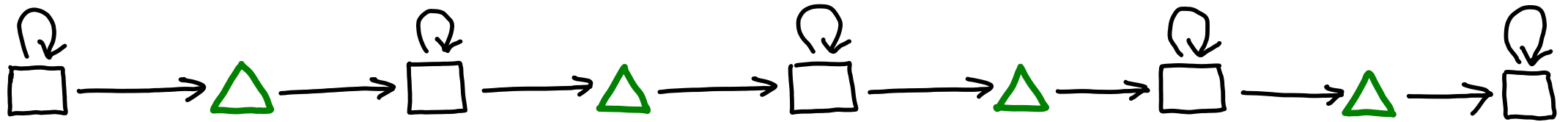
# A BÜCHI GAME



$attr_{\square}(T)$

$G' = W'_{\Delta}$   
 $attr_{\Delta}(W'_{\Delta})$

# A BÜCHI GAME



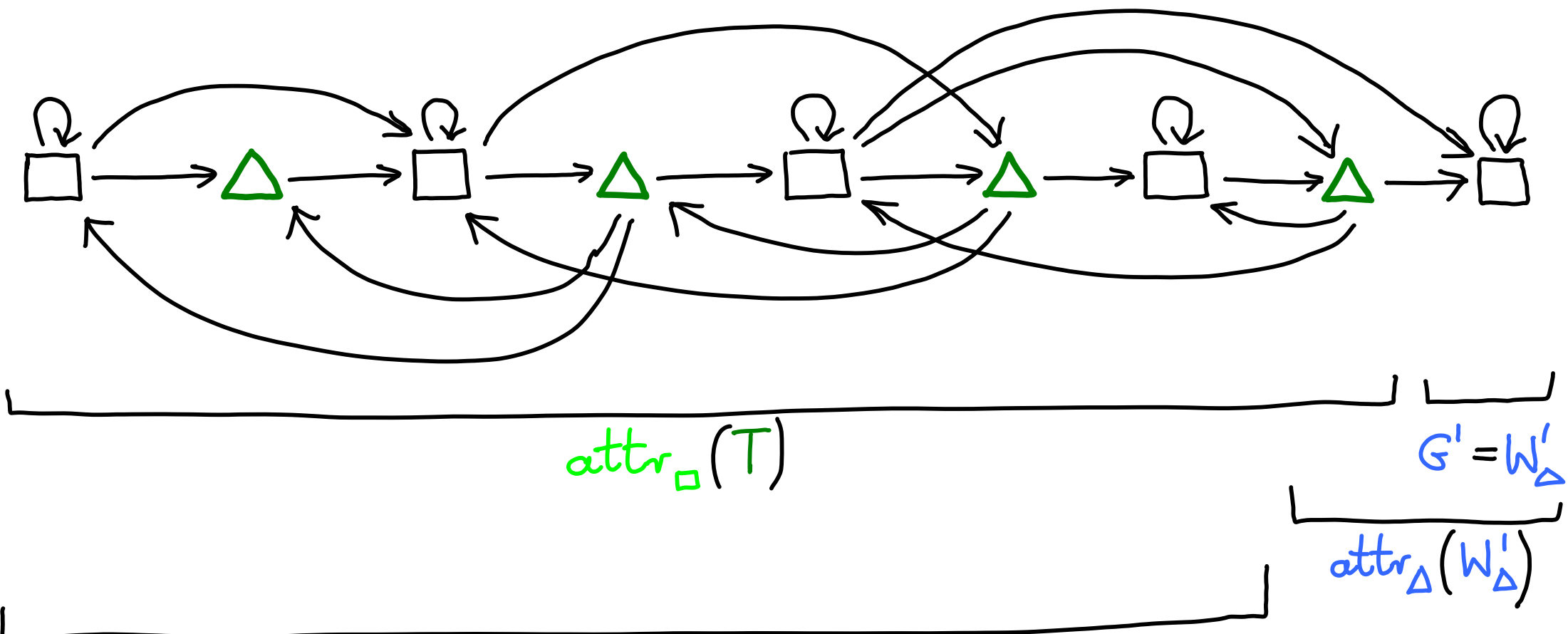
$attr_{\square}(T)$

$G' = W'_{\Delta}$



$G'' = G \setminus attr_{\Delta}(W'_{\Delta})$

# A BÜCHI GAME



$$G'' = G \setminus attr_{\Delta}(W'_{\Delta})$$

**FACT** The divide and conquer algorithm may take  $\Theta(mn)$  time



# AN $O(n^2)$ ALGORITHM FOR BÜCHI GAMES

[CHATTERJEE, HENZINGER 2012]

- Consider game graphs  $G_1, G_2, G_3, \dots, G_{\log n} = G$   
each  $G_i$  with  $\leq 2^i \cdot n$  edges, respectively
- Keep removing traps (complements of attractors)  
each time proceeding through graphs  $G_1, G_2, G_3, \dots, G_{\log n}$   
until one of them gives a non-empty trap

---

KEY PROPERTY If a trap found in  $G_i$  first  
then its size is  $\geq 2^i$

- Ammortized analysis charges  $(2 + 2^2 + \dots + 2^i) n \leq 2^{i+1} \cdot n$  work  
to the  $2^i$  vertices removed

# PLAN

1. Motivating example
2. Games on graphs
3. Reachability / safety games
4. Büchi /  $\omega$ -Büchi games
5. Parity games
6. Better algorithms for parity games
7. Search complexity of computing optimal strategies

# GAMES ON GRAPHS: PARITY OBJECTIVES

$$\text{Inf}(a_0, a_1, a_2, \dots) = \{a : a_i = a \text{ for infinitely many } i \in \mathbb{N}\}$$

Parity objective:

$$W_{\square} = \{(v_0, v_1, v_2, \dots) : \max(\text{Inf}(p(v_0), p(v_1), p(v_2), \dots)) \text{ is even}\}$$

where  $p: V \rightarrow \mathbb{N}$  is the priority function

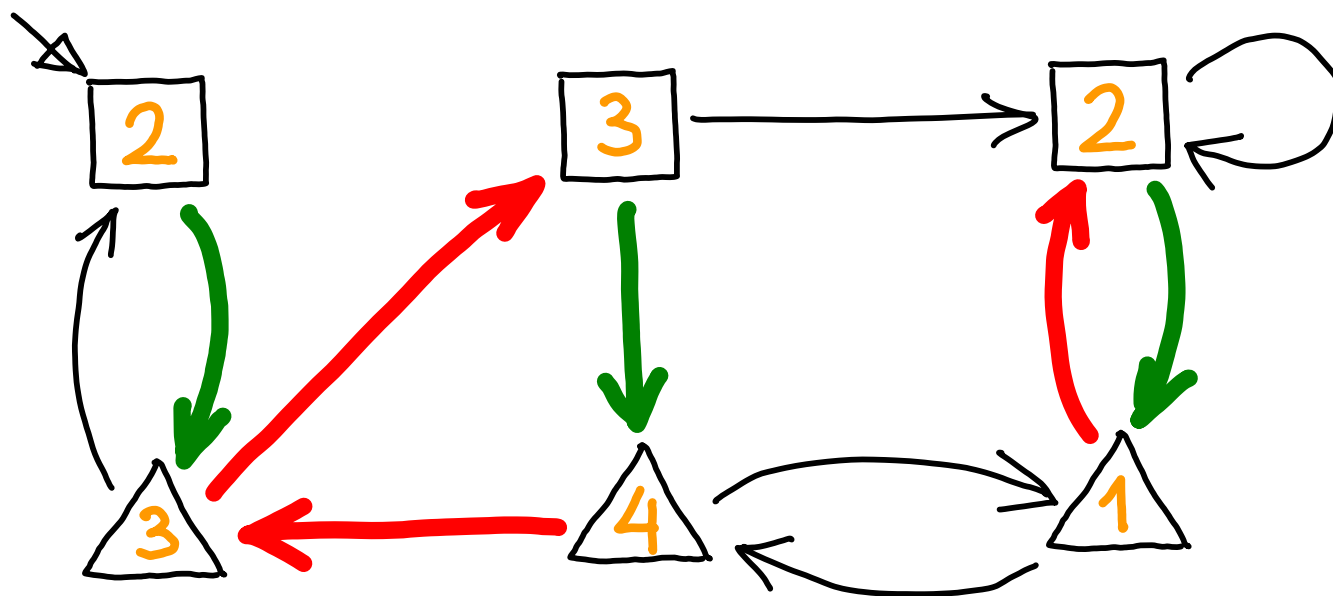
# PARITY GAMES

Players:

$6$  vertices,  $11$  edges,  $4$  priorities

Even =  $\square = 0$

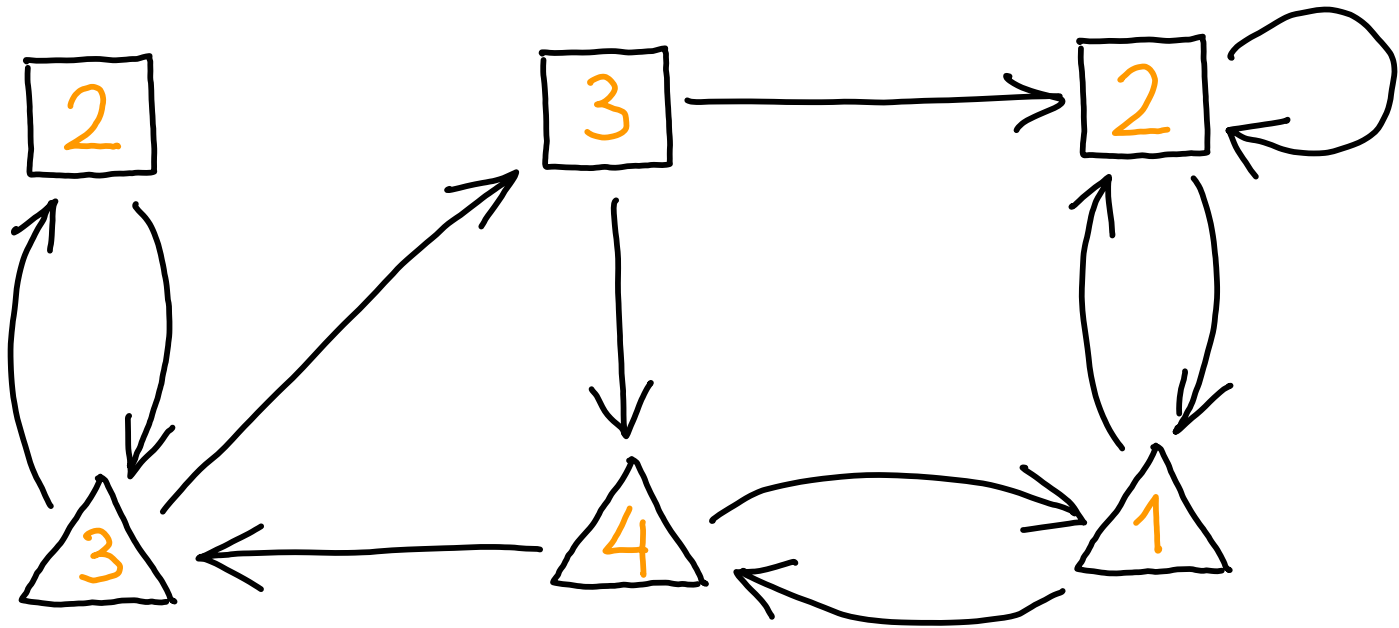
Odd =  $\triangle = 1$



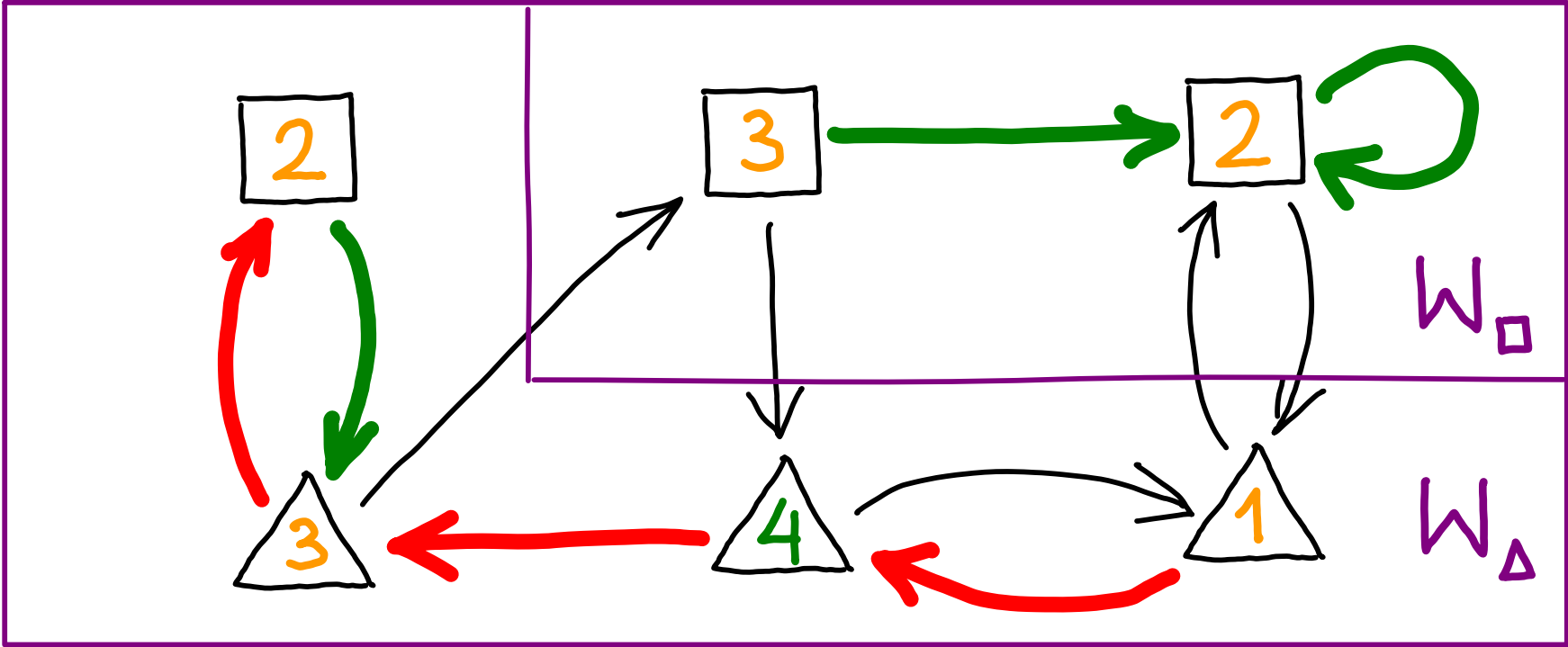
Winner of an infinite play:

parity of the highest priority occurring infinitely often

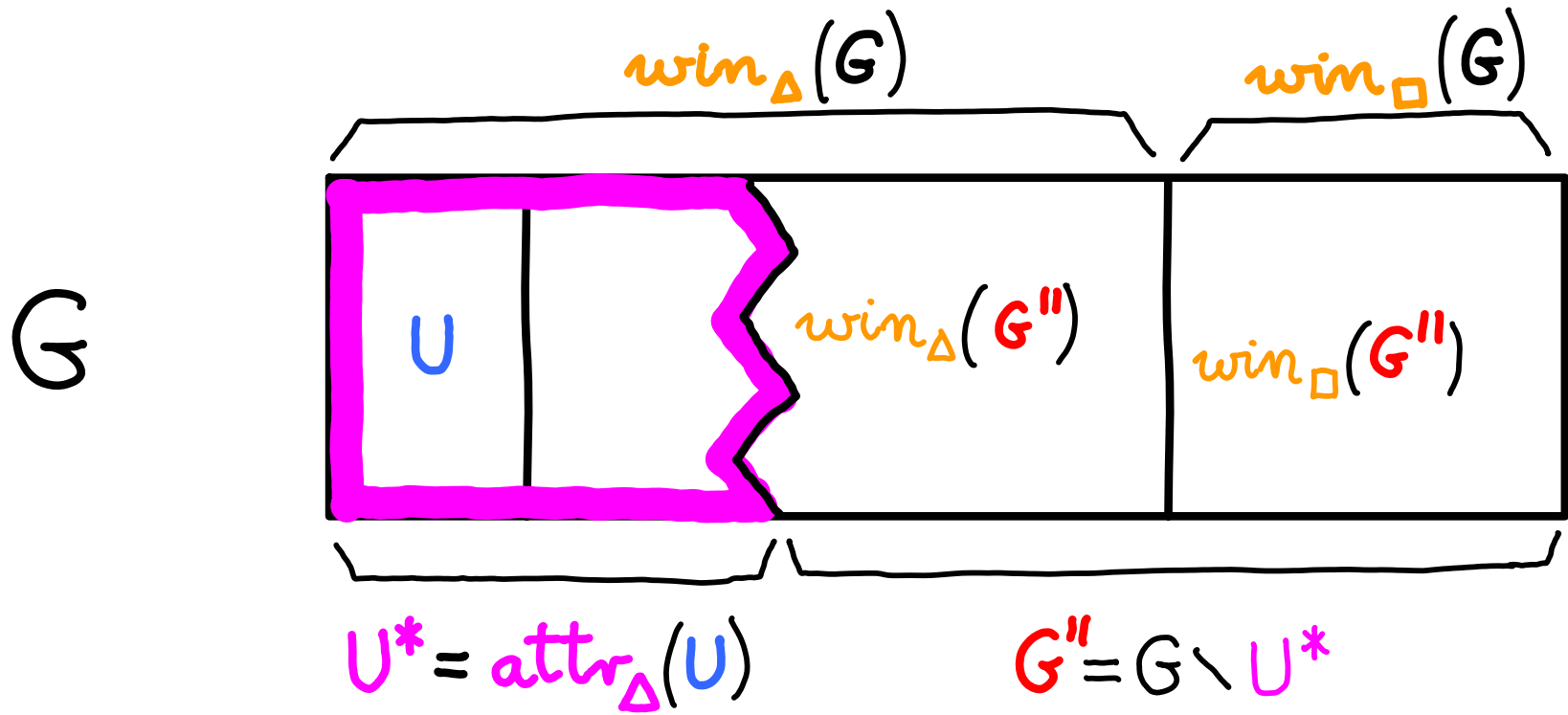
# SOLVING PARITY GAMES



# SOLVING PARITY GAMES



# ORACLE LEMMA



## LEMMA

If  $U \subseteq win_{\Delta}(G)$

then  $win_{\Delta}(G) = U^* \cup win_{\Delta}(G'')$

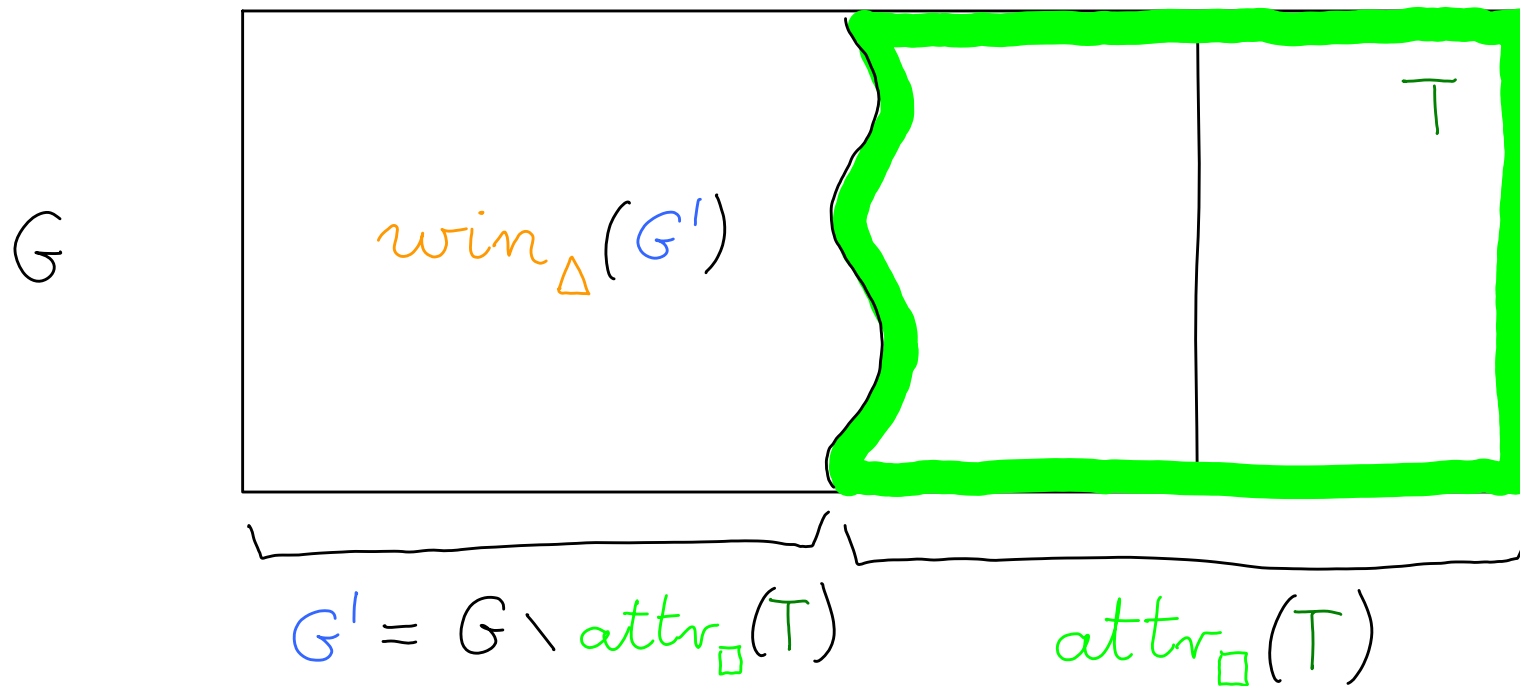
$win_{\square}(G) = win_{\square}(G'')$

## EXERCISE 5

- Does your proof of the oracle lemma for Büchi/co-Büchi games work also for parity games?
- Are the winning strategies constructed positional?
- If not, what assumptions do you need to make them so?



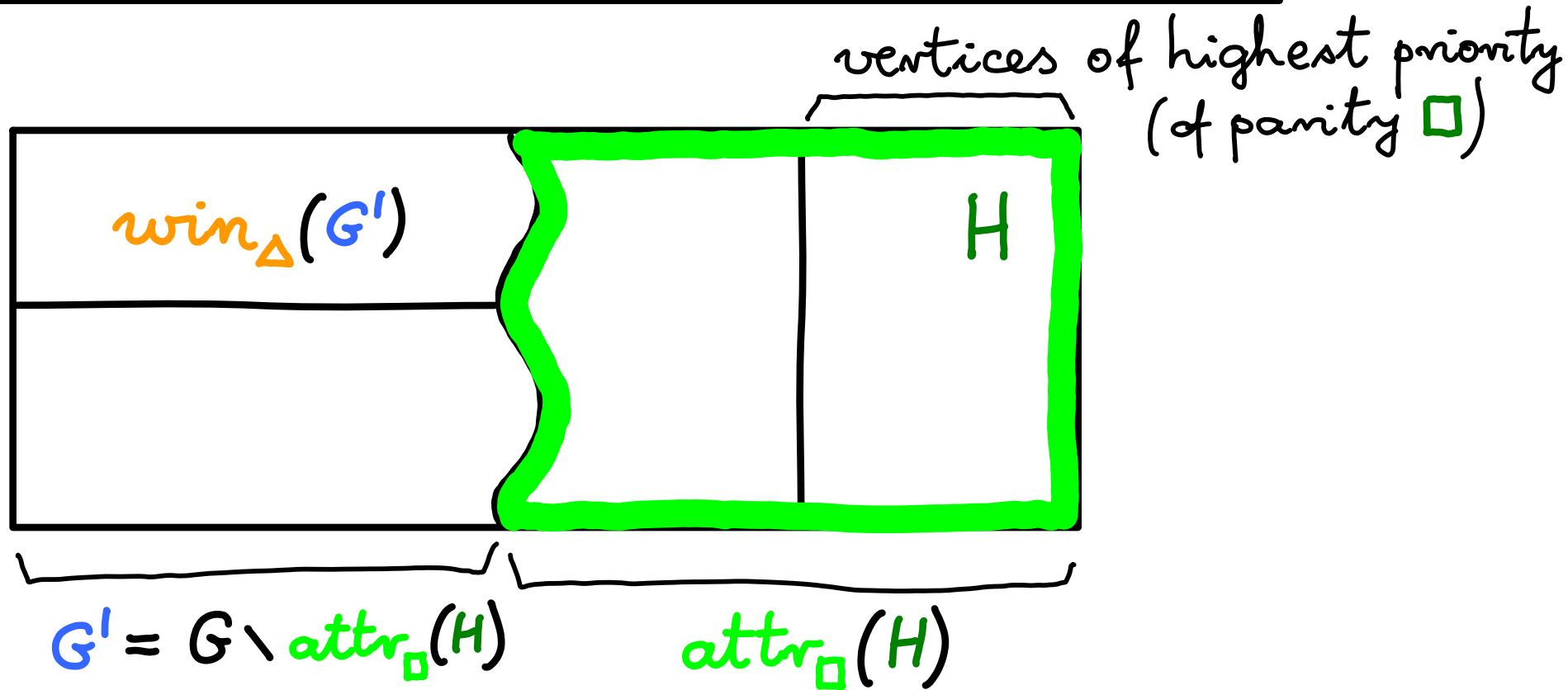
# T-ATTRACTOR LEMMA



## LEMMA

1.  $win_{\Delta}(G') = V \setminus attr_{\square}(T) \subseteq win_{\Delta}(G)$

# HIGHEST-PRIORITY ATTRACTOR LEMMA



## LEMMA

1. If  $win_{\Delta}(G') \neq \emptyset$  then  $win_{\Delta}(G') \subseteq win_{\Delta}(G)$

# T-ATTRACTOR LEMMA

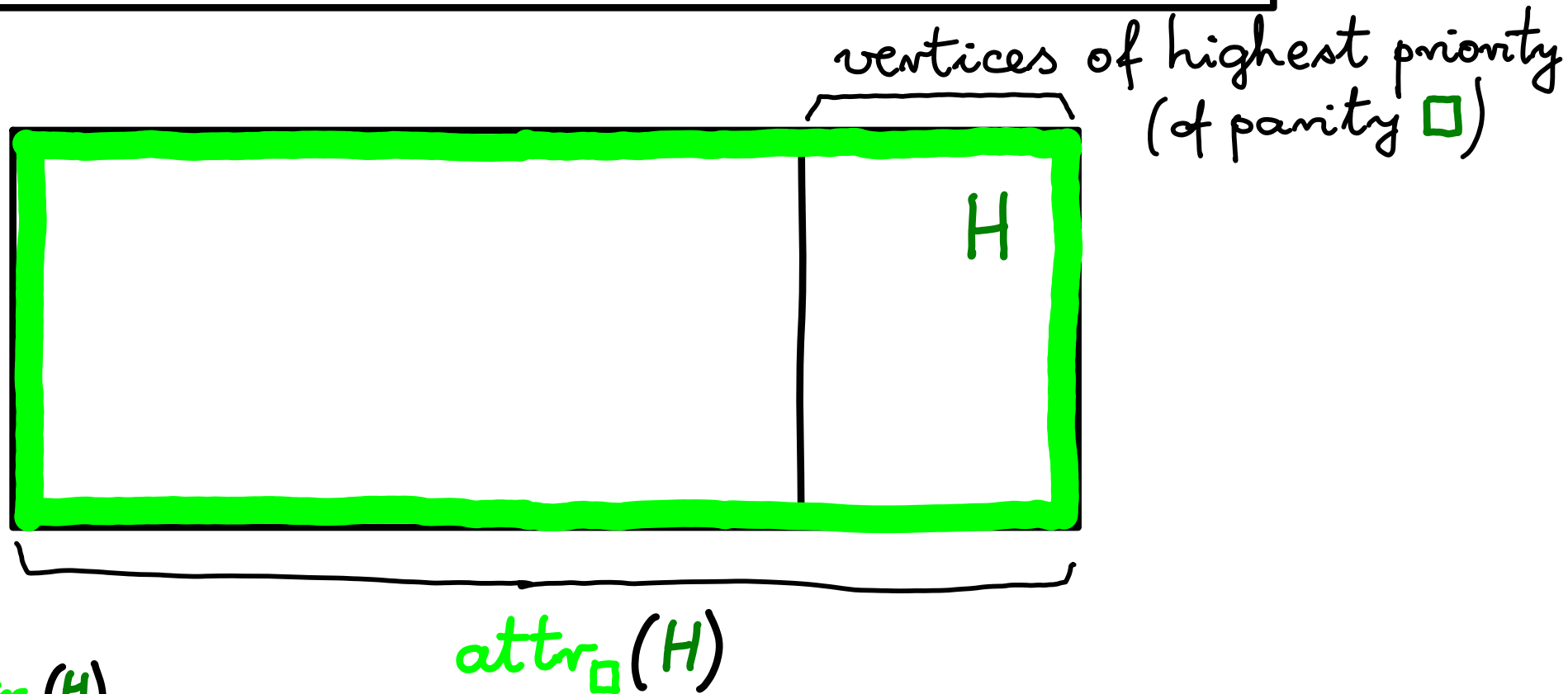


## LEMMA

1.  $win_{\Delta}(G') = V \setminus attr_{\square}(T) \subseteq win_{\Delta}(G)$

2. If  $G' = \emptyset$  then  $win_{\square}(G) = V$

# HIGHEST-PRIORITY ATTRACTOR LEMMA

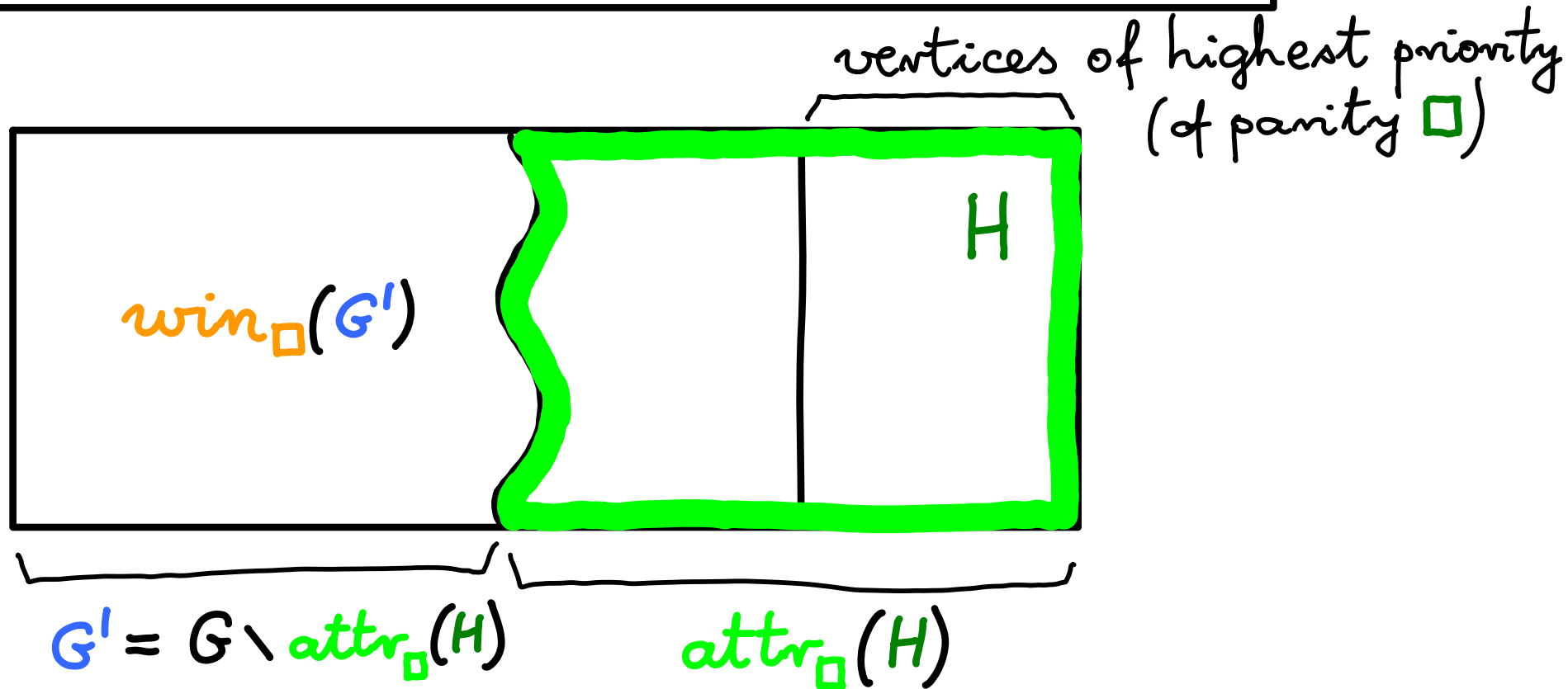


$$G' = G \setminus \text{attr}_{\square}(H)$$

## LEMMA

1. If  $\text{win}_{\Delta}(G') \neq \emptyset$  then  $\text{win}_{\Delta}(G') \subseteq \text{win}_{\Delta}(G)$
2. If  $G' = \emptyset$  then  $\text{win}_{\square}(G) = V$

# HIGHEST-PRIORITY ATTRACTOR LEMMA



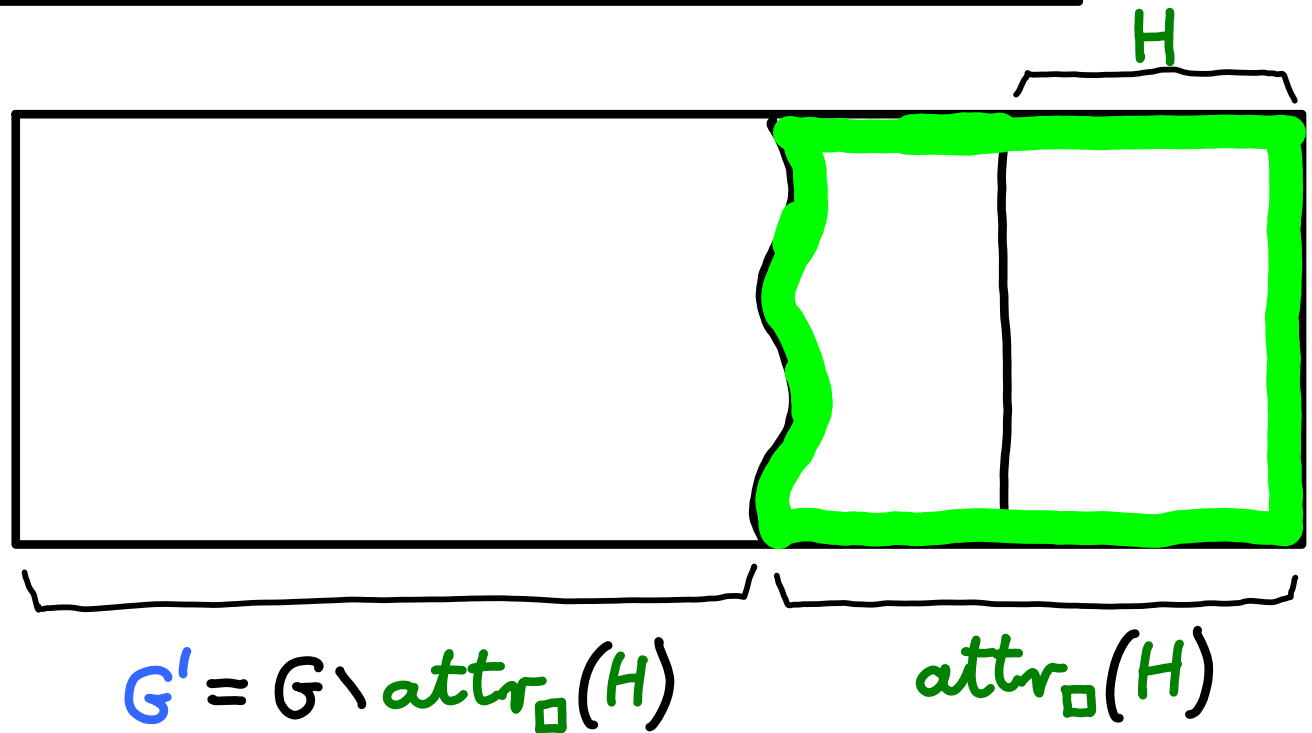
## LEMMA

1. If  $win_{\Delta}(G') \neq \emptyset$  then  $win_{\Delta}(G') \subseteq win_{\Delta}(G)$
2. If  $win_{\Delta}(G') = \emptyset$  then  $win_{\square}(G) = V$

## EXERCISE 6

Prove the highest-priority attractor lemma.

# A DIVIDE-AND-CONQUER ALGORITHM



WIN(G):  $(W_{\square}, W_{\Delta}) := \text{WIN}(G')$

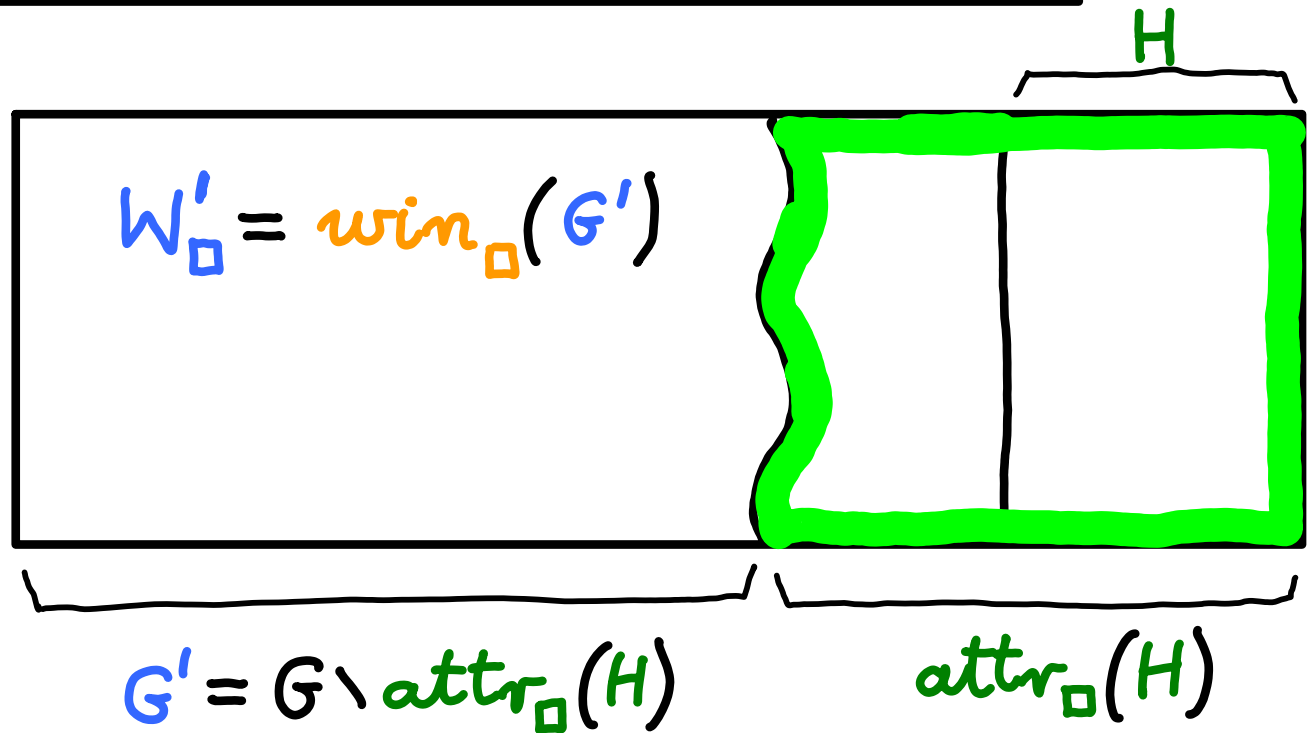
if  $W_{\Delta} = \emptyset$  then  $(W_{\square}, W_{\Delta}) := (V, \emptyset)$

else  $(W_{\square}, W_{\Delta}) := \text{WIN}(G'')$

$(W_{\square}, W_{\Delta}) := (W_{\square}, attr_{\Delta}(W'_{\Delta}) \cup W''_{\Delta})$

return  $(W_{\square}, W_{\Delta})$

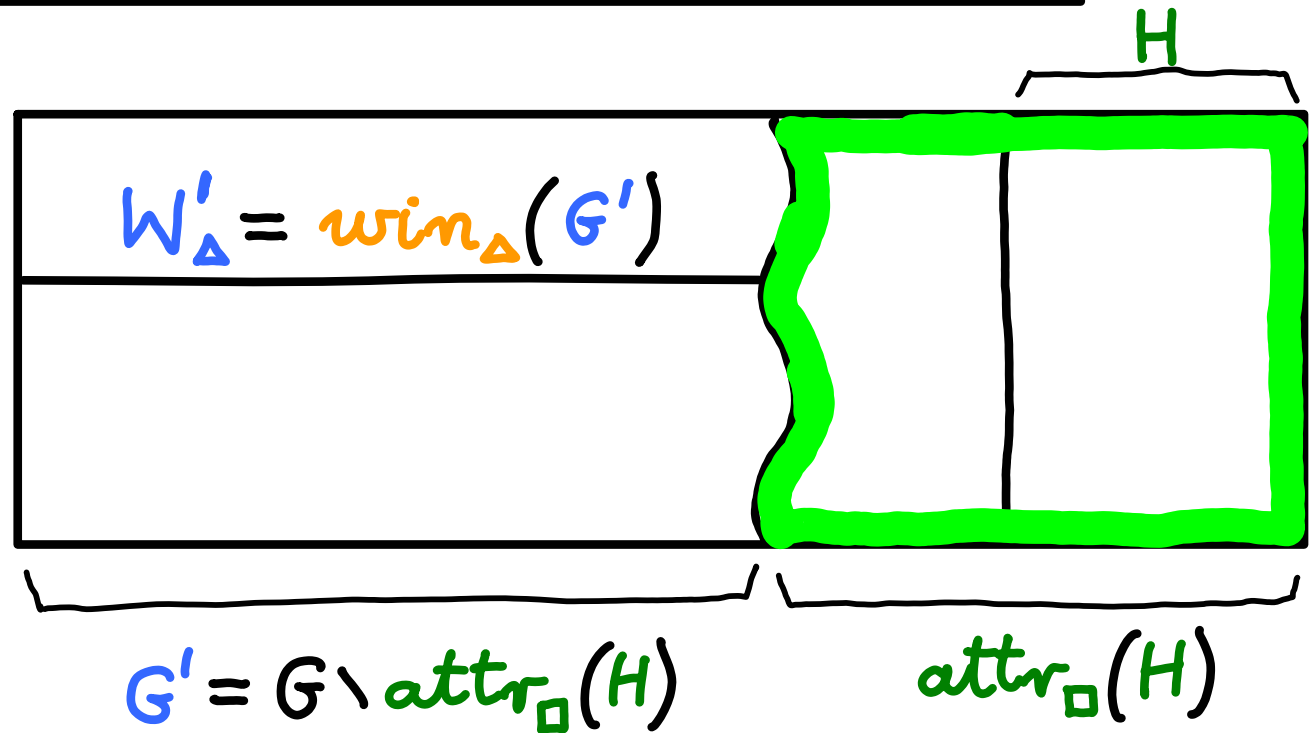
# A DIVIDE-AND-CONQUER ALGORITHM



WIN(G):  $(W_{\square}, W_{\Delta}) := \text{WIN}(G')$   
if  $W'_{\Delta} = \emptyset$  then  $(W_{\square}, W_{\Delta}) := (V, \emptyset)$   
else  $(W''_{\square}, W''_{\Delta}) := \text{WIN}(G'')$   
 $(W_{\square}, W_{\Delta}) := (W''_{\square}, \text{attr}_{\Delta}(W'_{\Delta}) \cup W''_{\Delta})$   
return  $(W_{\square}, W_{\Delta})$

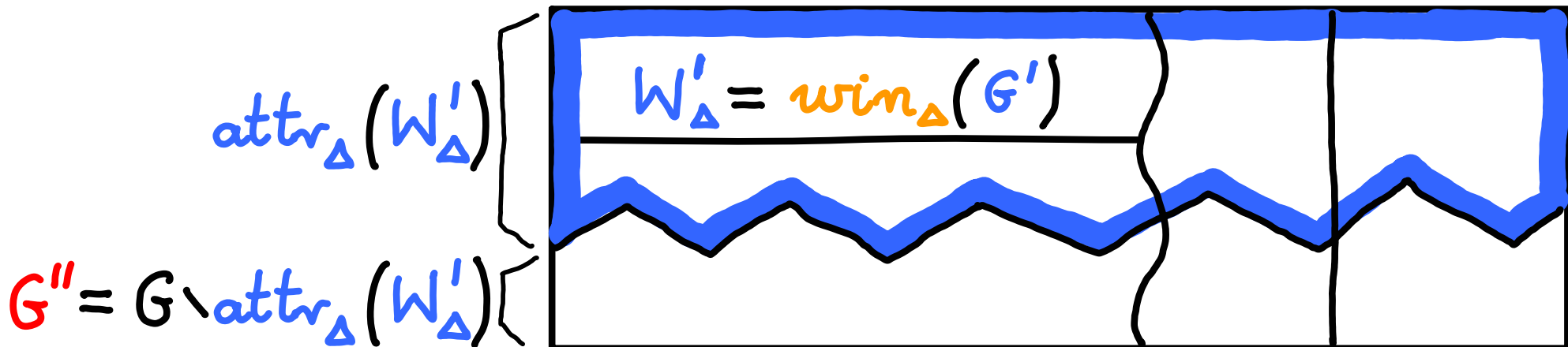


# A DIVIDE-AND-CONQUER ALGORITHM



WIN(G):  $(W_{\square}, W_{\Delta}) := WIN(G')$   
if  $W'_{\Delta} = \emptyset$  then  $(W_{\square}, W_{\Delta}) := (V, \emptyset)$   
else  $(W''_{\square}, W''_{\Delta}) := WIN(G'')$   
 $(W_{\square}, W_{\Delta}) := (W''_{\square}, attr_{\Delta}(W'_{\Delta}) \cup W''_{\Delta})$   
return  $(W_{\square}, W_{\Delta})$

# A DIVIDE-AND-CONQUER ALGORITHM

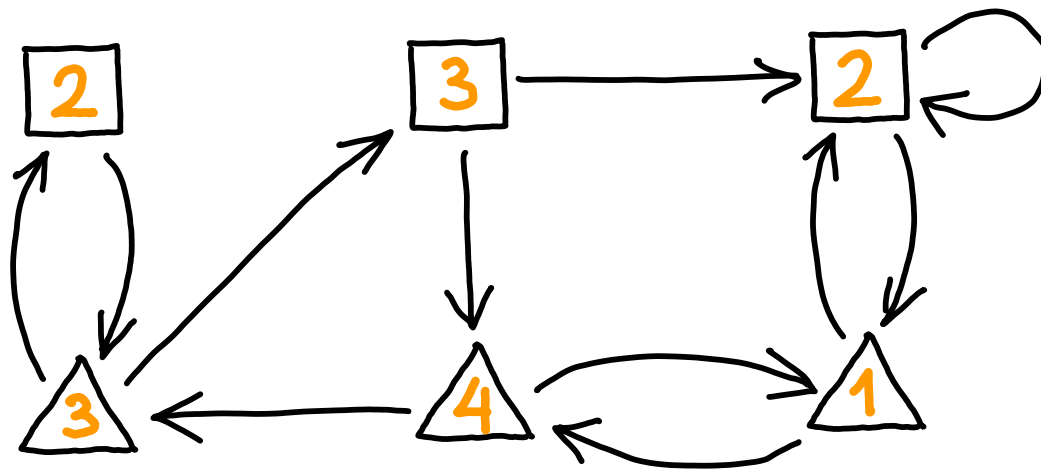


WIN(G):  $(W_\square, W_\Delta) := WIN(G')$   
if  $W'_\Delta = \emptyset$  then  $(W_\square, W_\Delta) := (V, \emptyset)$   
else  $(W''_\square, W''_\Delta) := WIN(G'')$   
 $(W_\square, W_\Delta) := (W''_\square, attr_\Delta(W'_\Delta) \cup W''_\Delta)$   
return  $(W_\square, W_\Delta)$

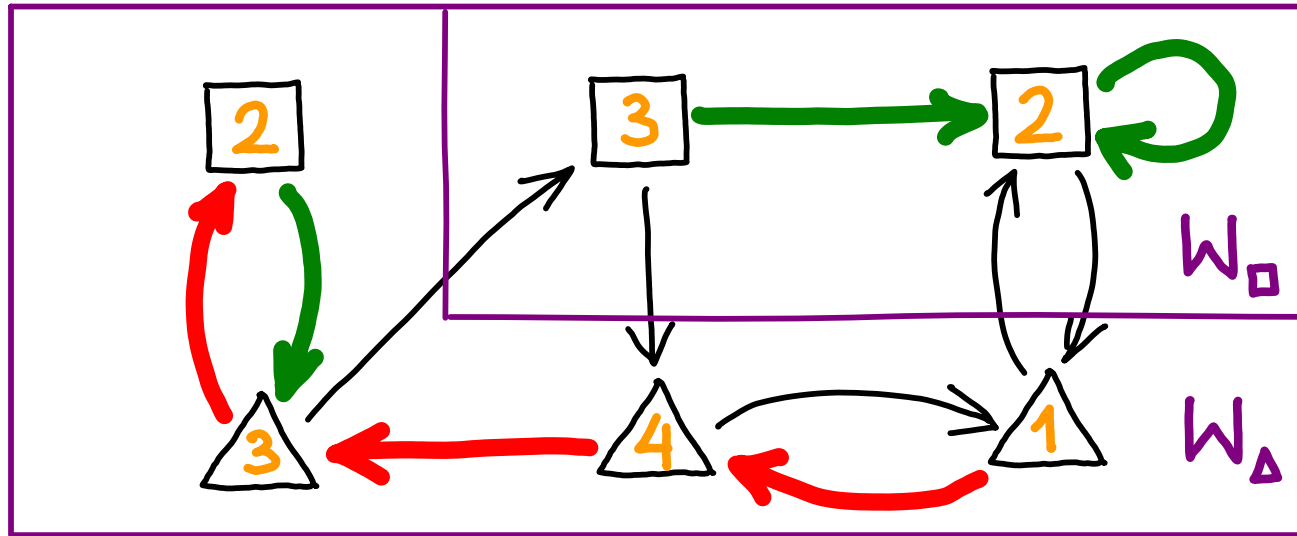
# EXERCISE 7

Run the divide-and-conquer algorithm on the following parity game.

Indicate the winning sets for both players, and their positional winning strategies.



# POSITIONAL DETERMINACY



THM [Emerson, Jutla; Mostowski 1991]

Parity games are positionally determined

## COROLLARY

(Deciding the winner in)  
parity games is in  $NP \cap co-NP$

## EXERCISE 8

- Use the oracle and highest-priority attractor lemmas, and induction on the size of the game graph, to prove the positional determinacy for parity games.
- Design a polynomial-time algorithm for checking if a given positional strategy of a player in a parity game is a winning strategy for her.

# RUNNING TIME OF THE DIVIDE-AND-CONQUER ALGORITHM

$T(n)$

WIN(G):  $(W'_0, W'_\Delta) := \text{WIN}(G')$   
if  $W'_\Delta = \emptyset$  then  $(W_0, W_\Delta) := (V, \emptyset)$   
else  $(W''_0, W''_\Delta) := \text{WIN}(G'')$   
 $(W_0, W_\Delta) := (W''_0, \text{attr}_\Delta(W'_\Delta) \cup W''_\Delta)$   
return  $(W_0, W_\Delta)$

$T(n-1)$

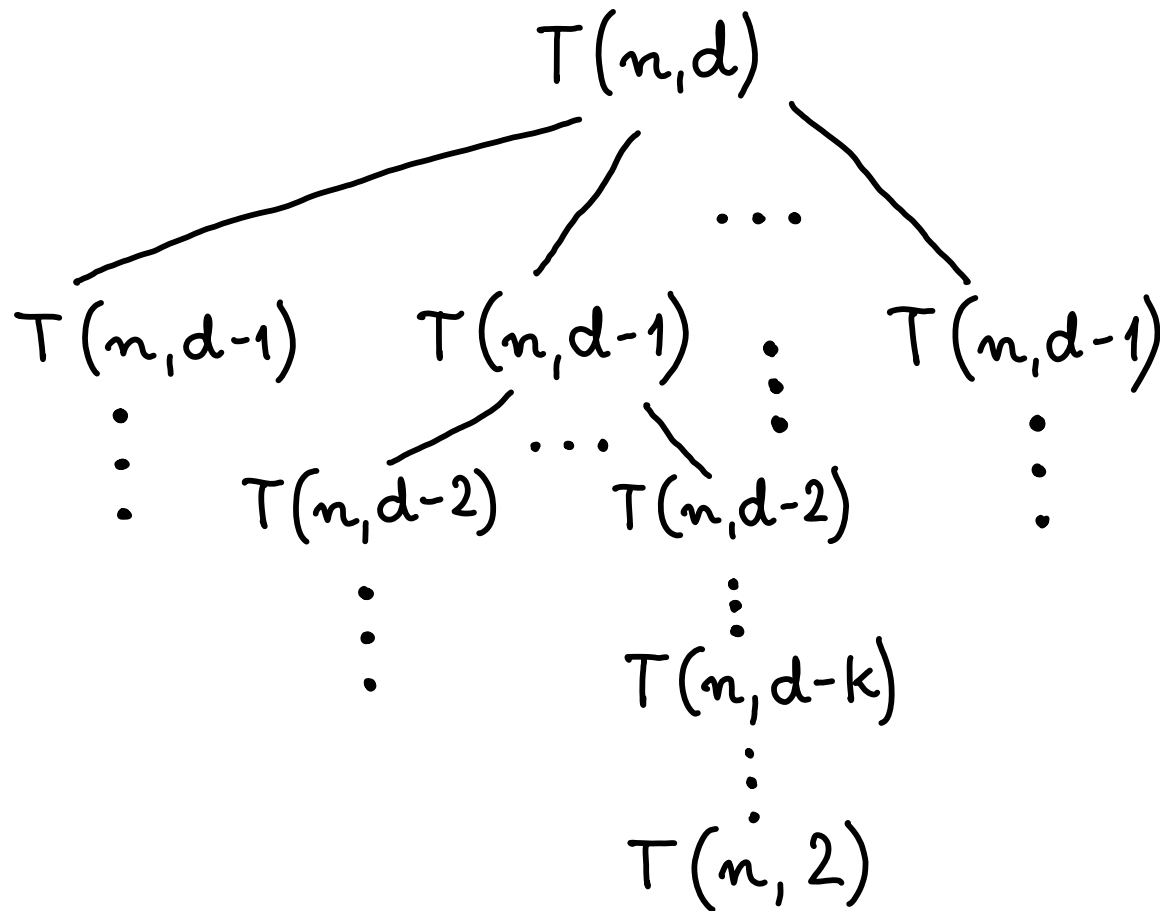
$T(n-1)$

Recurrence:  $T(n) \leq 2 \cdot T(n-1) + O(n^2)$

Solution:  $T(n) = O(2^n)$

# RUNNING TIME OF THE DIVIDE-AND-CONQUER ALGORITHM

The recursion tree



# children      Work per child

|           |          |
|-----------|----------|
| $n$       | $n^2$    |
| $n^2$     | $n^2$    |
| $n^3$     | $n^2$    |
| $\vdots$  | $\vdots$ |
| $n^{k+1}$ | $n^2$    |
| $\vdots$  | $\vdots$ |
| $n^{d-1}$ | $n^2$    |

---

$O(n^{d+O(1)})$

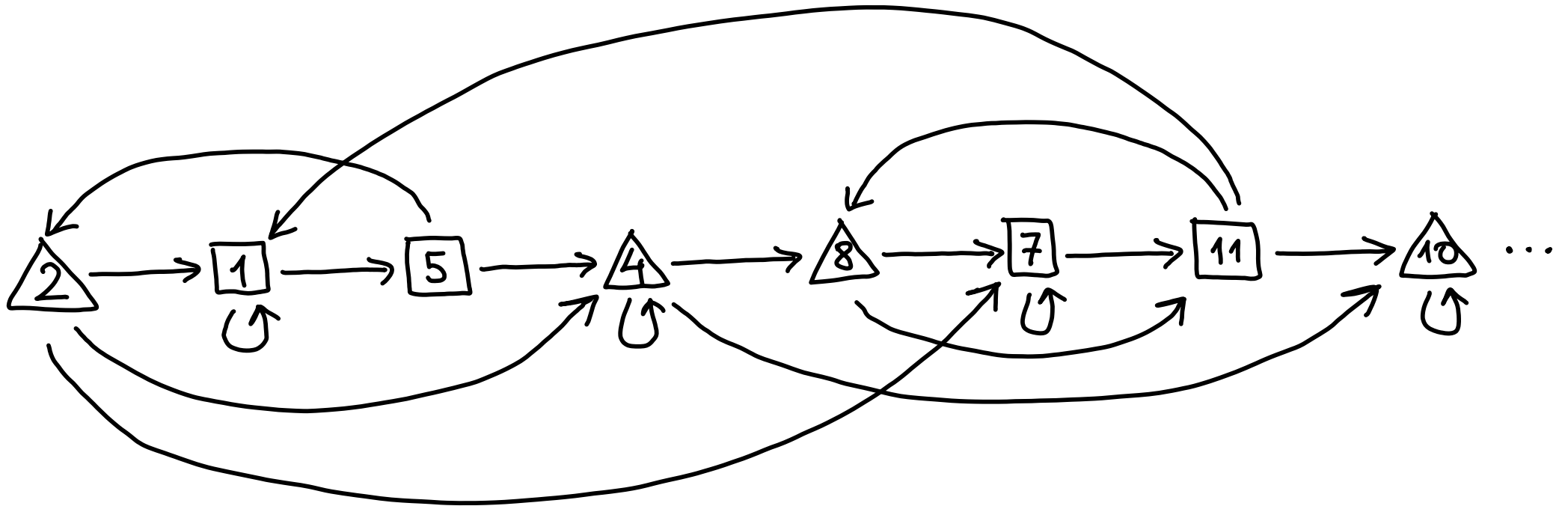
## EXERCISE 9

- Observe that the first recursive call  $\text{WIN}(G')$  is on the game  $G'$  with less than  $d$  distinct priorities.
- Argue that if the second recursive call  $\text{WIN}(G'')$  occurs then the game  $G''$  has less vertices of priority  $d$  than  $G$ .
- Conclude that the following recurrence:
$$\begin{cases} T(n, 1) = O(n) \\ T(n, d) \leq (n_d + 1) \cdot T(n, d-1) + O(n_d \cdot m) \end{cases}$$
characterizes the running time of the algorithm.
- Prove that  $T(n, d) = O\left(m \cdot \left(\frac{n+d}{d}\right)^d\right) = O\left(n^{d+O(1)}\right)$



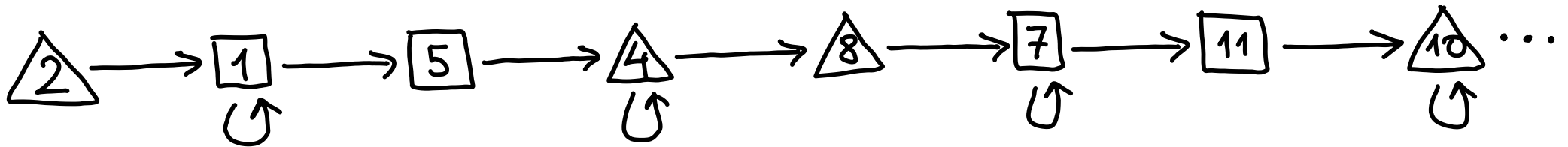
## EXERCISE 10

What is the running time of the divide-and-conquer algorithm on this parity game?



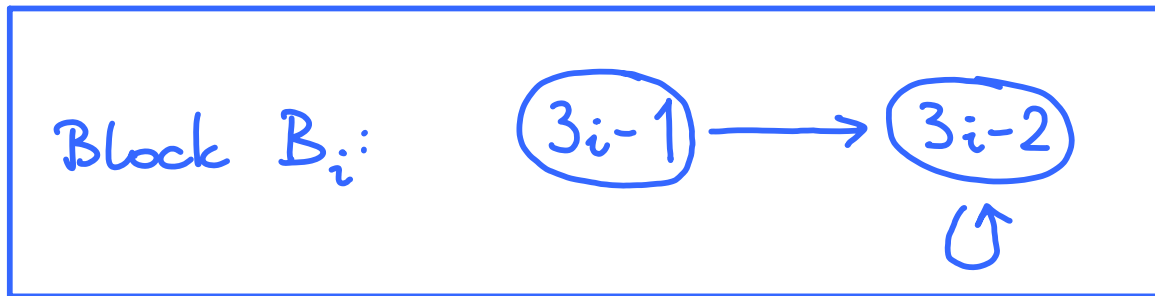
## EXERCISE 10

What is the running time of the divide-and-conquer algorithm on this parity game?

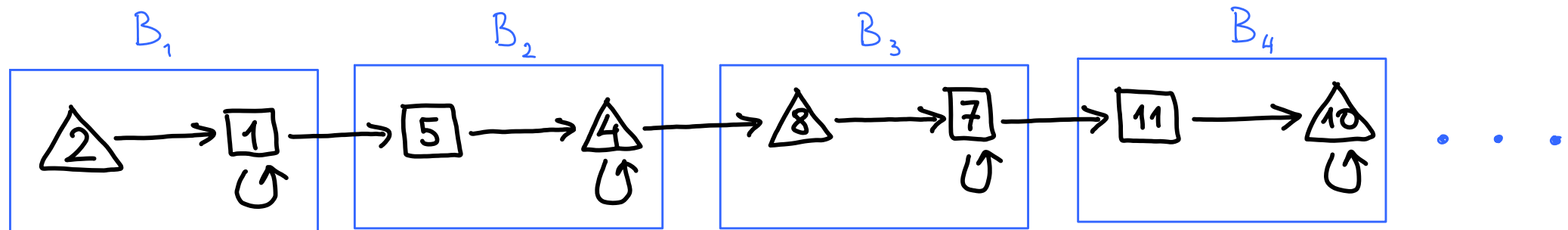


# EXERCISE 10

What is the running time of the divide-and-conquer algorithm on this parity game?



$\triangle i$  if  $i$  is even  
 $\square i$  if  $i$  is odd



# PLAN

1. Motivating example
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4. Büchi /  $\omega$ -Büchi games
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# BETTER ALGORITHMS

Many priorities

$$d = \Omega(n^{\frac{1}{2} + \epsilon})$$

Few priorities

$$d = O(n^{1/2})$$

Randomized

Expected

$$n^{O(\sqrt{n})}$$

RANDOM FACET

Deterministic

$$n^{O(\sqrt{n})}$$

[J., PATERSON,  
ZWICK 2008]

$$O(n^{\frac{d}{2} + O(1)})$$

[BROWNE ET AL. 1997, [SCHEWE 2007]  
SEIDL 1996,  
J. 2000]

$$O(n^{\frac{d}{3} + O(1)})$$

# BETTER ALGORITHMS

Many priorities

$$d = \Omega(n^{\frac{1}{2} + \epsilon})$$

Few priorities

$$d = O(n^{1/2})$$

Randomized

Expected

$$n^{O(\sqrt{n})}$$

Deterministic

$$n^{O(\sqrt{n})}$$

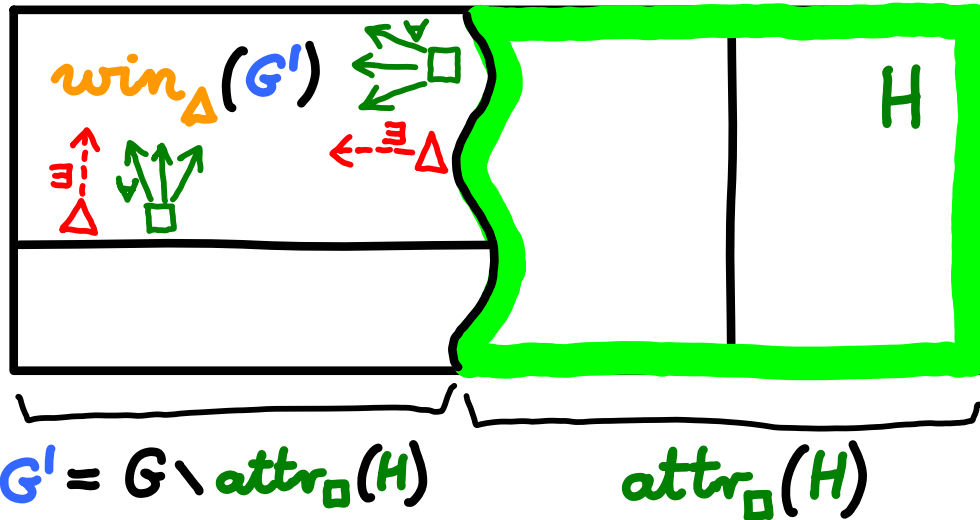
$$O(n^{\frac{d}{2} + O(1)})$$

$$O(n^{\frac{d}{3} + O(1)})$$

# DOMINIONS

DEF  $D \subseteq V$  is an **dominion** for  $\Delta$  if  $\Delta$  has a trapping strategy in  $D$  that is winning for her

---



FACT  $win_{\Delta}(G')$  is a dominion for  $\Delta$  in  $G$

---

FACT A dominion of size  $\leq l$  can be found in  $n^{o(l)}$  time

## EXERCISE 11

Design a brute-force algorithm for detecting dominions of size  $\leq \ell$  that runs in time  $O(n^\ell)$ .



# A SUBEXPONENTIAL ALGORITHM

$T(n)$

WIN(G): If there is a dominion  $D \neq \emptyset$  of size  $\leq \ell$   
 then return WIN(G \setminus D) \oplus D

else

$(W_{\square}, W_{\Delta}) := \text{WIN}(G')$

if  $W_{\Delta} = \emptyset$  then  $(W_{\square}, W_{\Delta}) := (V, \emptyset)$

else  $(W'_{\square}, W'_{\Delta}) := \text{WIN}(G'')$

$(W_{\square}, W_{\Delta}) := (W'_{\square}, \text{attr}_{\Delta}(W'_{\Delta}) \cup W'_{\Delta})$

return  $(W_{\square}, W_{\Delta})$

$n^{o(\ell)}$

$T(n-1)$

$T(n-1)$

$T(n-\ell)$

Recurrence:  $T(n) \leq n^{o(\ell)} + T(n-1) + T(n-\ell)$

Solution:

$$T(n) = n^{O(\sqrt{n})}$$

if  $\ell = \sqrt{n}$

## EXERCISE 12

- Define labelled tree  $\tau_n$ 
  - root labelled  $n$
  - if  $n > 3$  then the root has:
    - \* left subtree  $\tau_{n-1}$
    - \* right subtree  $\tau_{n-\lfloor\sqrt{2n}\rfloor}$
  - if  $n = 1, 2, \text{ or } 3$  then the root has no children
- Prove that on every path from root to leaf in  $\tau_n$  there are at most  $\lfloor\sqrt{2n}\rfloor$  right turns.

Hint: Use induction and the fact that

$$\text{if } \frac{1}{2}j^2 < n \leq \frac{1}{2}(j+1)^2 \text{ then } n - \lfloor\sqrt{2n}\rfloor \leq \frac{1}{2}j^2$$

- Conclude that  $\tau_n$  has  $n^{O(\sqrt{n})}$  leaves.

# BETTER ALGORITHMS

Many priorities

$$d = \Omega\left(n^{\frac{1}{2} + \epsilon}\right)$$

Few priorities

$$d = O\left(n^{1/2}\right)$$

Randomized

Expected

$$n^{O(\sqrt{n})}$$

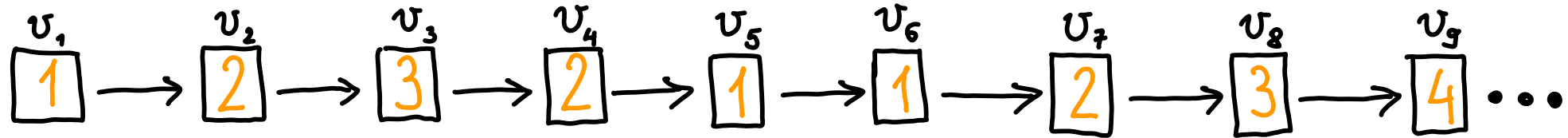
Deterministic

$$n^{O(\sqrt{n})}$$

$$O\left(n^{\frac{d}{2} + O(1)}\right)$$

$$O\left(n^{\frac{d}{3} + O(1)}\right)$$

# A SUFFICIENT CONDITION FOR WINNING PLAYS



$\mu_3: \quad 2 \geq 2 \geq 2 > 1 \geq 1 \geq 1 \geq 1 > 0$   
 $\mu_1: \quad 1 > 0 \quad \quad \quad 2 > 1 > 0$

## FACT

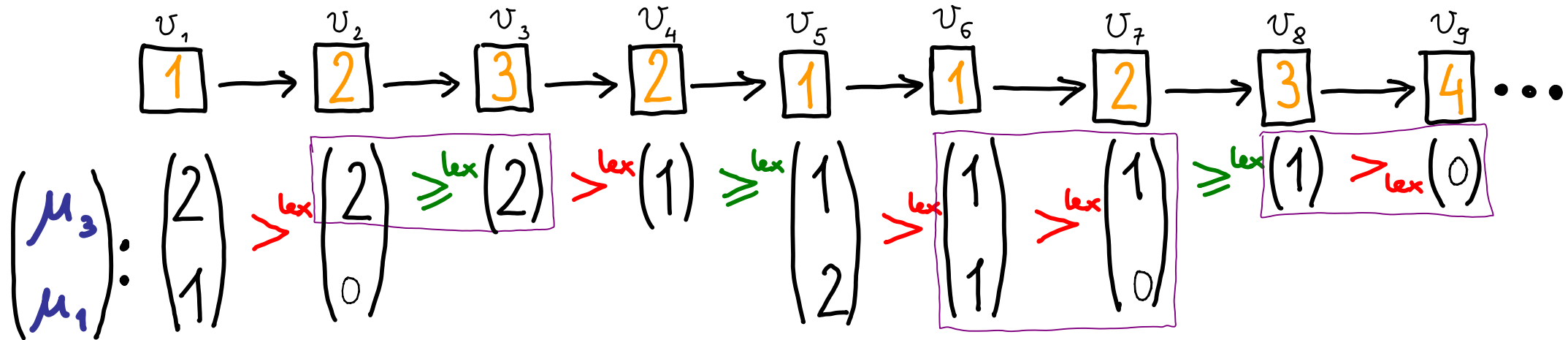
$v_1, v_2, v_3, \dots$  is winning for  $\square$

if

there are  $\mu_1, \mu_3, \dots, \mu_{d-1}: V \rightarrow \mathbb{N}$ , s.t.

- $\mu_k(v_i) \geq \mu_k(v_{i+1})$  for  $p(v_i) < k \leq d-1$
- $\mu_k(v_i) > \mu_k(v_{i+1})$  for  $k = p(v_i)$

# A SUFFICIENT CONDITION FOR WINNING PLAYS



## FACT

$v_1, v_2, v_3, \dots$  is winning for  $\square$

if

there are  $\mu_1, \mu_3, \dots, \mu_{d-1} : V \rightarrow \mathbb{N}$ , s.t.

$$\begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v_i)} \end{pmatrix} (v_i) \geq_{\text{lex}} \begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v_{i+1})} \end{pmatrix} (v_{i+1})$$

$>_{\text{lex}}$  if  $p(v_i)$  is odd!

## EXERCISE 13

- Prove the two facts.
- Argue that the following holds for the first characterization:  
if  $\mu_k(v) = l$   
then, starting from  $v$ ,  
at most  $l$  vertices of priority  $k$  will occur,  
before a vertex of priority  $> k$  occurs.
- Does a similar property hold for the second characterization?

# CHARACTERIZING EXISTENCE OF WINNING PLAYS

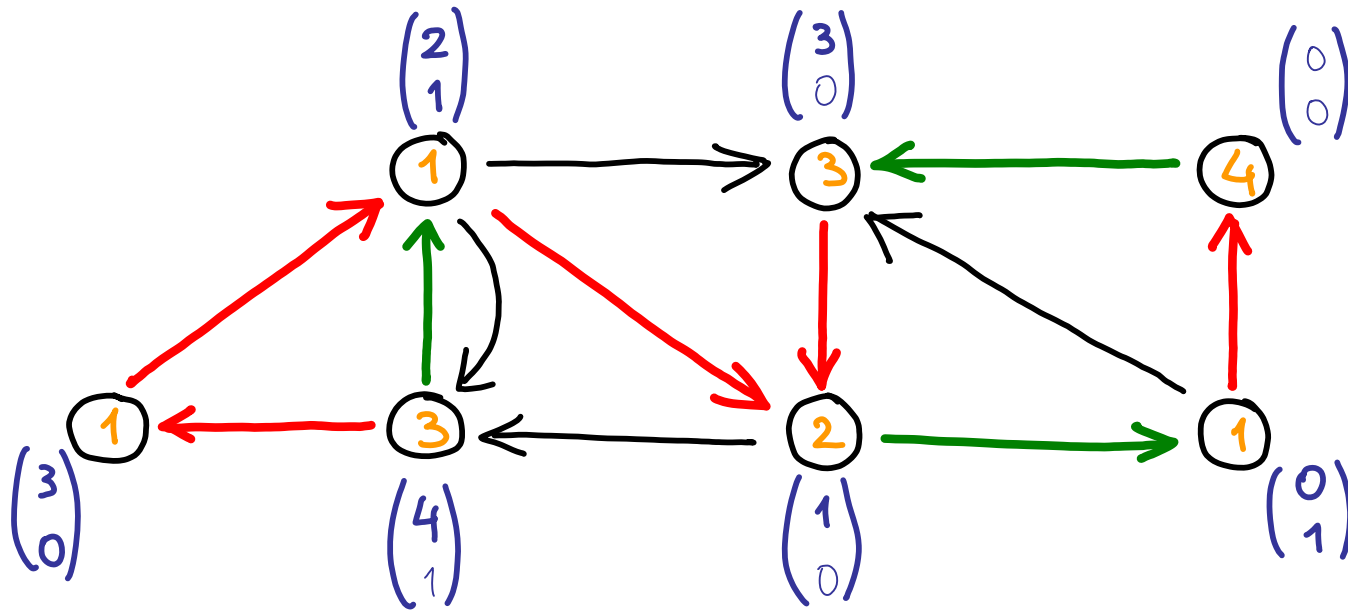
## LEMMA

There **is** a winning play (from every vertex) for  $\square$   
if there are  $\mu_1, \mu_3, \dots, \mu_{d-1}: V \rightarrow \mathbb{N}$ , s.t.

$$\begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{P(v)} \end{pmatrix} (v) \geq^{\text{lex}} \min_{(v,w) \in E} \begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{P(v)} \end{pmatrix} (w)$$

$>^{\text{lex}}$  if  $P(v)$   
is odd!

# CHARACTERIZING EXISTENCE OF WINNING PLAYS



$$\mu_3(v) >^{\text{lex}} \min_{(v,w) \in E} \mu_3(w) \quad \text{if } p(v) = 3$$

$$\mu_3(v) \geq^{\text{lex}} \min_{(v,w) \in E} \mu_3(w) \quad \text{if } p(v) = 2$$

$$\begin{pmatrix} \mu_3 \\ \mu_1 \end{pmatrix}(v) >^{\text{lex}} \min_{(v,w) \in E} \begin{pmatrix} \mu_3 \\ \mu_1 \end{pmatrix}(w) \quad \text{if } p(v) = 1$$



# CHARACTERIZING UNIVERSALITY OF WINNING PLAYS

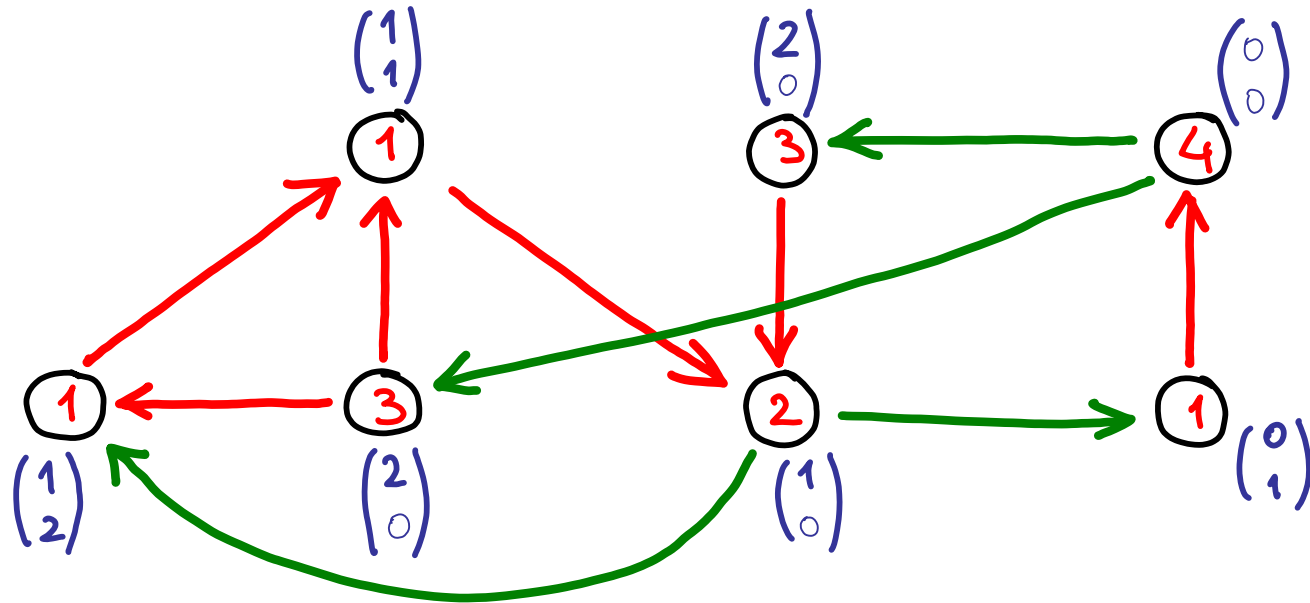
## LEMMA

**ALL** plays (from every vertex) are winning for  $\square$   
if there are  $\mu_1, \mu_3, \dots, \mu_{d-1}: V \rightarrow \mathbb{N}$ , s.t.

$$\begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (v) \geq^{\text{lex}} \max_{(v,w) \in E} \begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (w)$$

$>^{\text{lex}}$  if  $p(v)$   
is odd!

# CHARACTERIZING UNIVERSALITY OF WINNING PLAYS



$$\mu_3(v) >^{\text{lex}} \max_{(v,w) \in E} \mu_3(w) \quad \text{if } p(v) = 3$$

$$\mu_3(v) \geq^{\text{lex}} \max_{(v,w) \in E} \mu_3(w) \quad \text{if } p(v) = 2$$

$$\begin{pmatrix} \mu_3 \\ \mu_1 \end{pmatrix}(v) >^{\text{lex}} \max_{(v,w) \in E} \begin{pmatrix} \mu_3 \\ \mu_1 \end{pmatrix}(w) \quad \text{if } p(v) = 1$$

# CHARACTERIZING EXISTENCE OF WINNING STRATEGIES

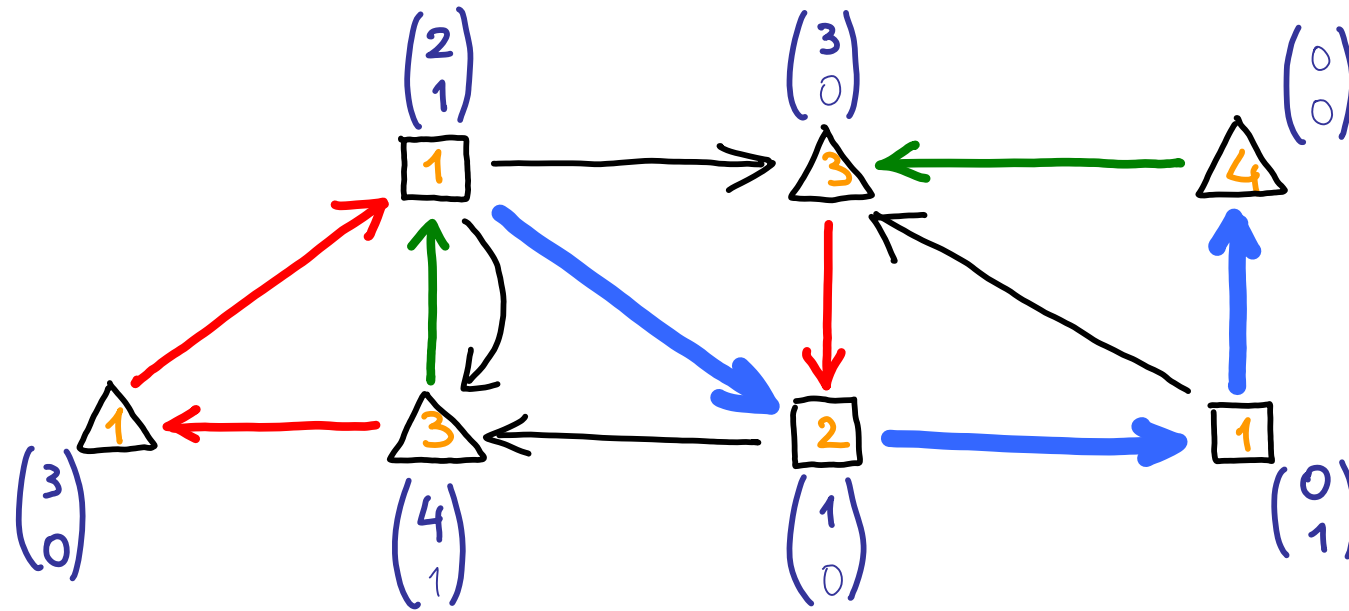
## THEOREM

There is a winning strategy for  $\square$  (from every vertex)  
iff there are  $\mu_1, \mu_3, \dots, \mu_{d-1}: V \rightarrow \mathbb{N}$ , s.t.

- $$\begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (v) \geq^{\text{lex}} \min_{(v,w) \in E} \begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (w) \quad \begin{matrix} >^{\text{lex}} & \text{if } p(v) \\ & \text{is odd!} \end{matrix} \quad \text{for } v \in V_{\square}$$

- $$\begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (v) \geq^{\text{lex}} \max_{(v,w) \in E} \begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (w) \quad \begin{matrix} >^{\text{lex}} & \text{if } p(v) \\ & \text{is odd!} \end{matrix} \quad \text{for } v \in V_{\Delta}$$

# CHARACTERIZING EXISTENCE OF WINNING STRATEGIES



$$v \in V_{\square}$$

$$v \in V_{\triangle}$$

$$\mu_3(v) >^{lex} \min_{(v,w) \in E} \mu_3(w)$$

$$\mu_3(v) >^{lex} \max_{(v,w) \in E} \mu_3(w)$$

$$\text{if } p(v) = 3$$

$$\mu_3(v) \geq^{lex} \min_{(v,w) \in E} \mu_3(w)$$

$$\mu_3(v) \geq^{lex} \max_{(v,w) \in E} \mu_3(w)$$

$$\text{if } p(v) = 2$$

$$\begin{pmatrix} \mu_3 \\ \mu_1 \end{pmatrix}(v) >^{lex} \min_{(v,w) \in E} \begin{pmatrix} \mu_3 \\ \mu_1 \end{pmatrix}(w)$$

$$\begin{pmatrix} \mu_3 \\ \mu_1 \end{pmatrix}(v) >^{lex} \max_{(v,w) \in E} \begin{pmatrix} \mu_3 \\ \mu_1 \end{pmatrix}(w)$$

$$\text{if } p(v) = 1$$

# SMALL PROGRESS MEASURES

## THEOREM

There is a winning strategy for  $\square$  (from every vertex)  
iff there are  $\mu_1, \mu_3, \dots, \mu_{d-1}: V \rightarrow \mathbb{N}$ , s.t.

- $$\begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (v) \geq^{\text{lex}} \min_{(v,w) \in E} \begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (w)$$

$>^{\text{lex}}$  if  $p(v)$  is odd!

for  $v \in V_{\square}$

- $$\begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (v) \geq^{\text{lex}} \max_{(v,w) \in E} \begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_{p(v)} \end{pmatrix} (w)$$

$>^{\text{lex}}$  if  $p(v)$  is odd!

for  $v \in V_{\Delta}$

and  $\mu_k: V \rightarrow \{0, 1, 2, \dots, n_k\}$  where  $n_k = |p^{-1}(k)|$

## EXERCISE 14

Prove the small progress measure theorem.

Hint: Use positional determinacy of parity games and induction on the size of the winning set similar to that in the proof of correctness of the divide-and-conquer algorithm.

# SMALL PROGRESS MEASURES

## LEMMA

The number of tuples

$$\begin{pmatrix} \mu_{d-1} \\ \vdots \\ \mu_1 \end{pmatrix} \in \mathbb{N}^{d/2}$$

s.t.  $\mu_k \in \{0, 1, 2, \dots, n_k\}$

where  $n_k = |p^{-1}(k)|$

is  $\leq \left(\frac{n}{d/2}\right)^{d/2}$

$$= O\left(n^{\frac{d}{2}}\right)$$

# PROGRESS MEASURE LIFTING ALGORITHM

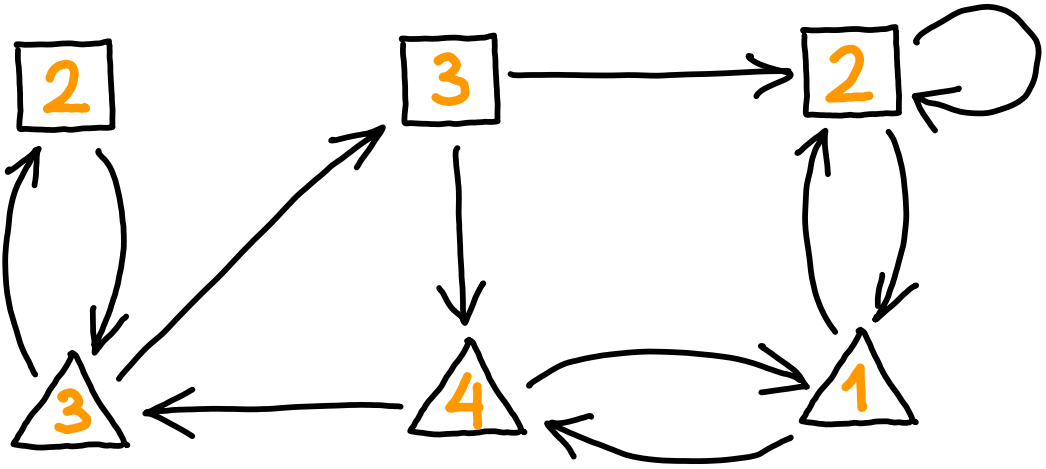
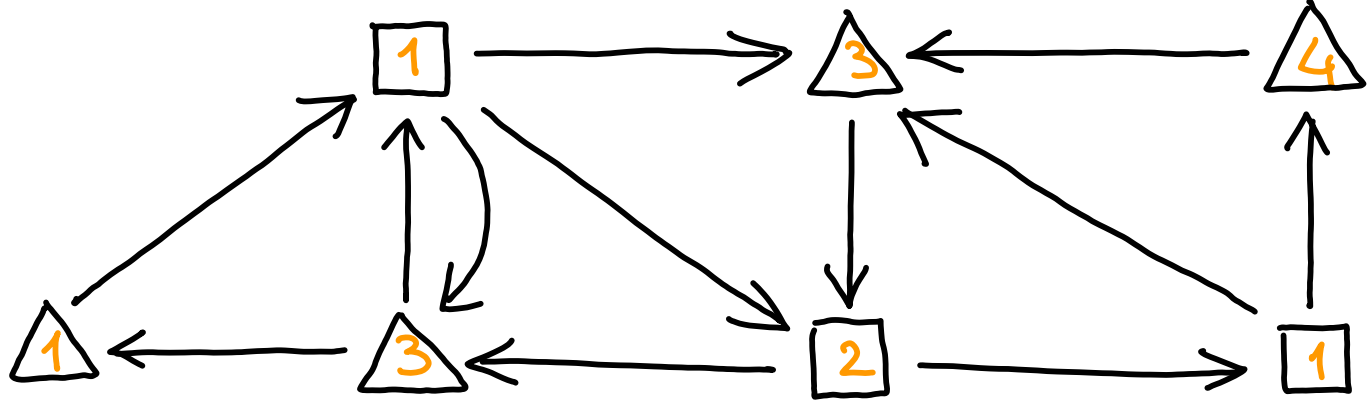
1. Start with  $\mu: V \ni v \mapsto (0, 0, \dots, 0)$
2. While  $\mu$  violates the progress inequality at some  $v \in V$ , lift  $\mu(v)$  (minimally) so that it doesn't

Running time:  $\leq dm \cdot n^{\frac{d}{2}}$   
 $= O\left(n^{\frac{d}{2} + O(1)}\right)$



# EXERCISE 15

Run the progress-measure-lifting algorithm on the following parity games:



# BETTER ALGORITHMS

Many priorities

$$d = \Omega\left(n^{\frac{1}{2} + \epsilon}\right)$$

Few priorities

$$d = O\left(n^{1/2}\right)$$

Randomized

Expected

$$n^{O(\sqrt{n})}$$

Deterministic

$$n^{O(\sqrt{n})}$$

$$O\left(n^{\frac{d}{2} + O(1)}\right)$$

$$O\left(n^{\frac{d}{3} + O(1)}\right)$$

# SMALLER PROGRESS MEASURES FOR DOMINIONS

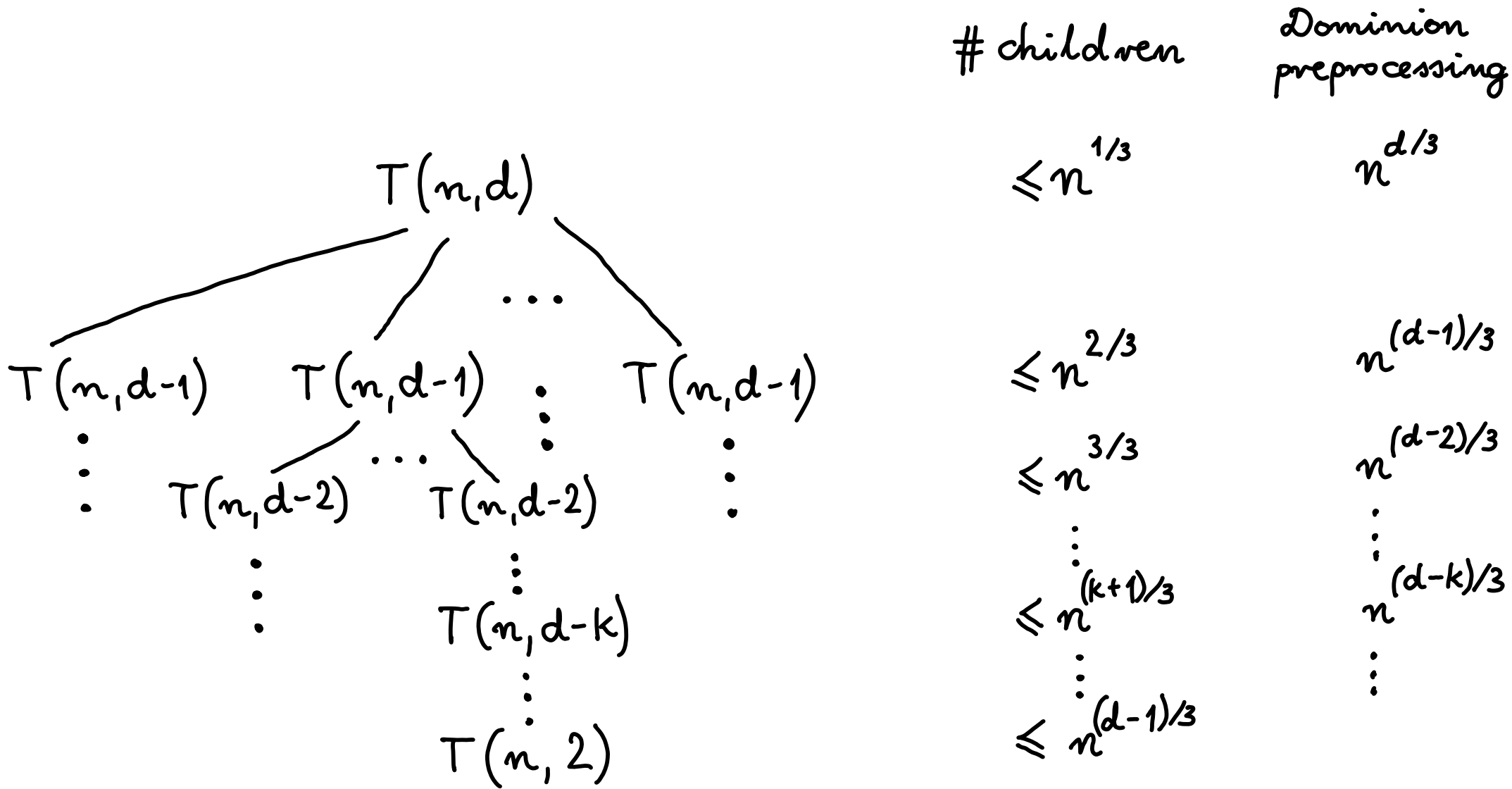
THM If  $D \subseteq V$  is a dominion of size  $\leq \ell$ ,  
then there is a progress measure  $\mu: V \rightarrow M_G^\infty(\ell)$ ,  
s.t.  $\mu(D) \subseteq M_G(\ell)$ ,

where  $M_G(\ell) = \left\{ \mu \in \mathbb{N}^{d/2} : \sum_{k=1}^{d/2} \mu_k \leq \ell \right\}$

COR

1. All dominions of size  $\leq \ell/2$   
can be found in time  $O\left(\ell + \frac{d}{2}\right)$
2. All dominions of size  $\leq n^{2/3}$   
can be found in time  $O(n^{d/3})$

# THE RECURSION TREE FOR SCHEWE'S ALGORITHM



---


$$O(n^{\frac{d}{3}} + O(1))$$

## EXERCISE 16

- Prove that

$$\underbrace{\left| \left\{ \mu \in \mathbb{N}^{d/2} : \sum_{k=1}^{d/2} \mu_k \leq \ell \right\} \right|}_{M_G(\ell)} = \binom{\ell + \frac{d}{2}}{\frac{d}{2}} = \prod_{i=1}^{d/2} \frac{\ell + i}{i}$$

- Argue that if  $\ell = 2 \cdot \lceil n^{2/3} \rceil$

then  $|M_G(\ell)| \leq n^{d/3}$  for all  $d \geq 8$ .

- Prove that if the dominion returned by the progress measure lifting algorithm, using  $M_G(\ell)$ , is of size  $\leq \ell/2$  then there is no other dominion of size  $\leq \ell/2$  in  $G$ .

## EXERCISE 17

- Argue that the running time of Schewé's algorithm can be characterized by the recurrence:

$$\begin{cases} T(n, 1) = O(n) \\ T(n, d) \leq T(n, d-1) + T(n - \lceil n^{2/3} \rceil, d) + O(n^{d/3 + O(1)}) \end{cases}$$

and hence also by:

$$\begin{cases} T(n, 1) = O(n) \\ T(n, d) \leq n^{1/3} \cdot T(n, d-1) + O(n^{d/3 + O(1)}) \end{cases}$$

- Prove that  $T(n, d) = O(n^{d/3 + O(1)})$ .

## EXERCISE 18

- Give algorithms with the smallest asymptotic worst-case running time bound as you can find, for parity games with 2, 3, 4, 5, 6, 7, and 8 priorities.

# PLAN

1. Motivating example
2. Games on graphs
3. Reachability / safety games
4. Büchi /  $\omega$ -Büchi games
5. Parity games
6. Better algorithms for parity games
7. Search complexity of computing optimal strategies



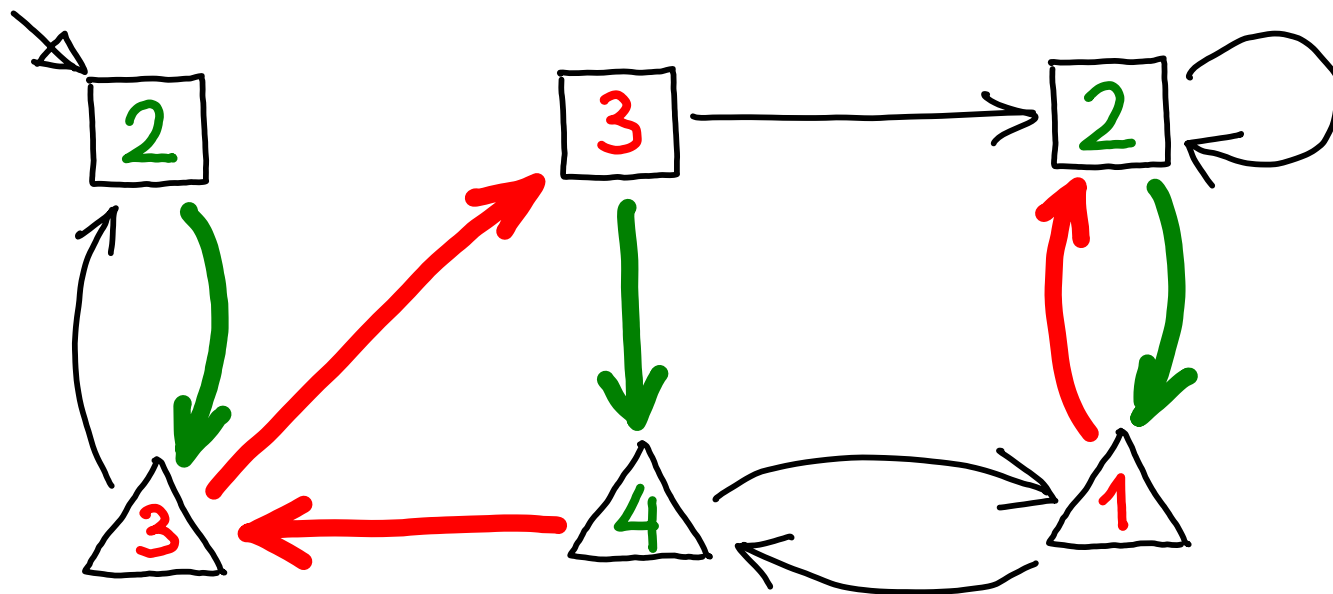
# PARITY GAMES

Players:

$n$  vertices,  $m$  edges,  $d$  priorities

Even =  $\square = 0$

Odd =  $\triangle = 1$



Winner of an infinite play:

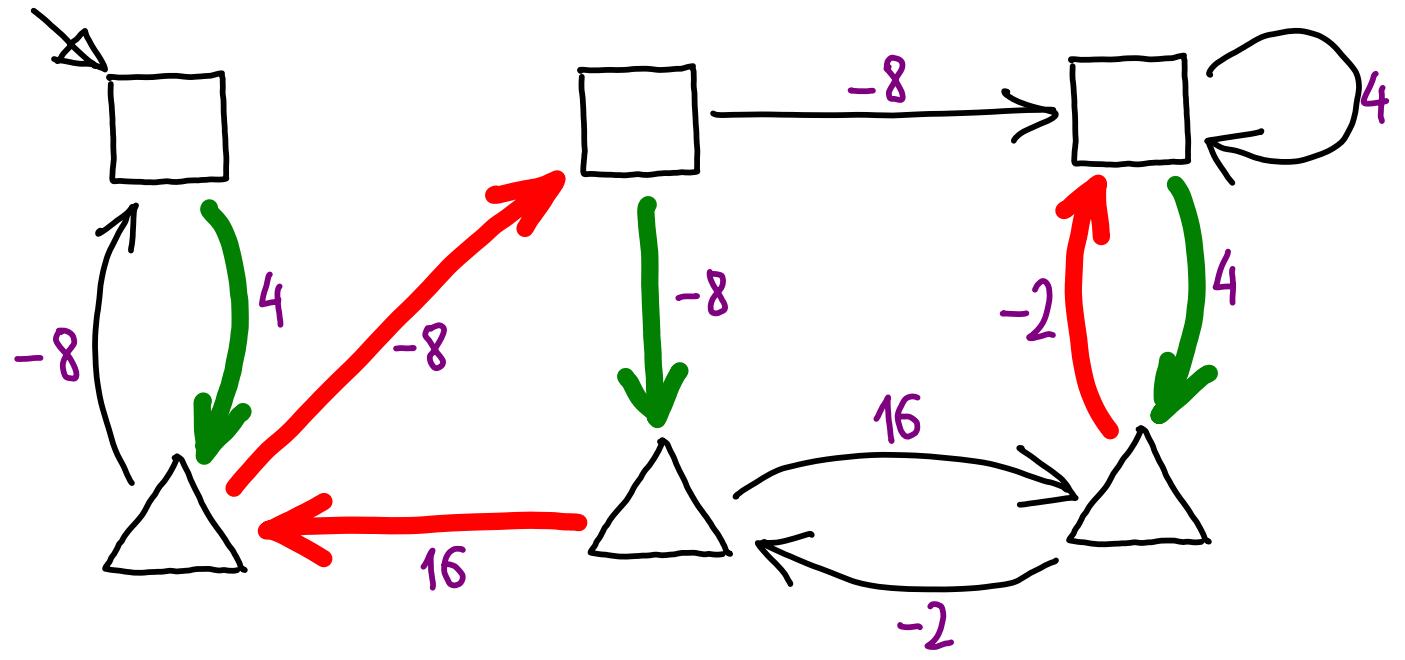
parity of the highest priority occurring infinitely often

# MEAN-PAYOFF GAMES

Players:

MAX = □

MIN = △



Winner of an infinite play  $\pi = \langle v_0, v_1, v_2, \dots \rangle$

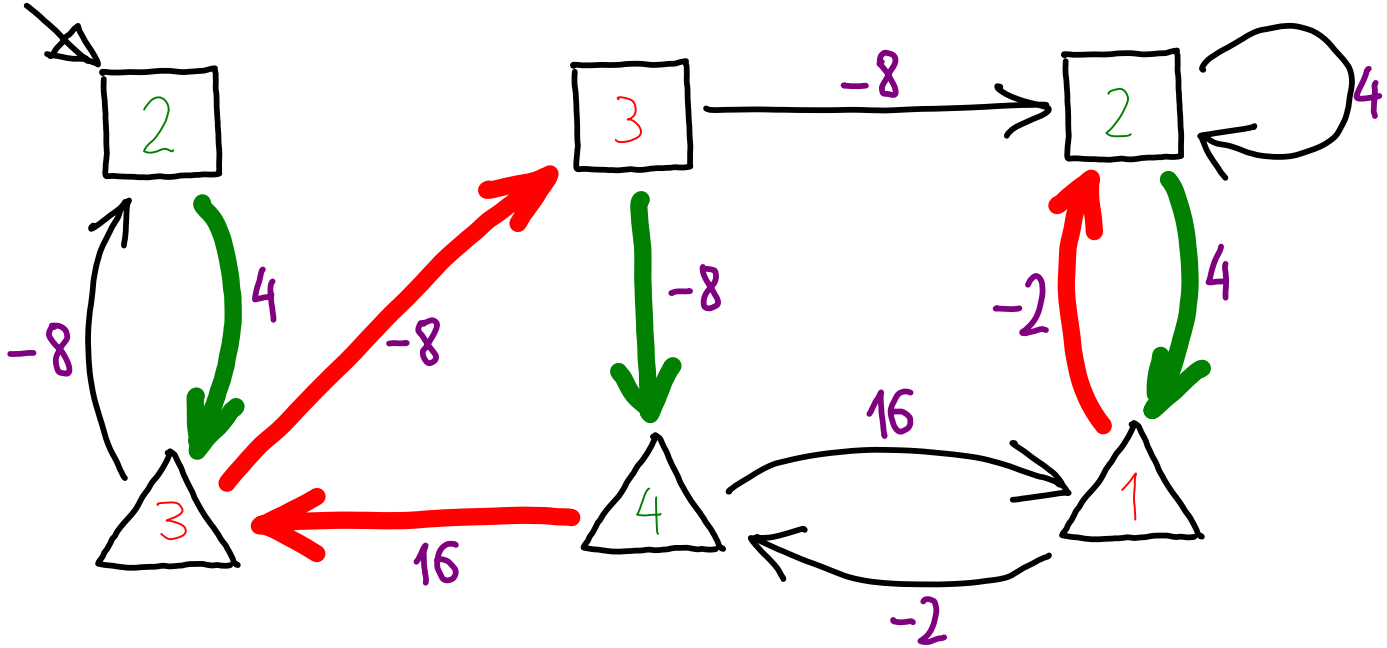
MAX if  $\lim_{n \rightarrow \infty} \left( \frac{1}{n} \cdot \sum_{i=0}^{n-1} r(v_i, v_{i+1}) \right) \geq 0$

# MEAN-PAYOFF GAMES

Players:

MAX = □

MIN = △



$$(v, w) \longmapsto (-n)^{p(v)}$$

here 2 was good enough

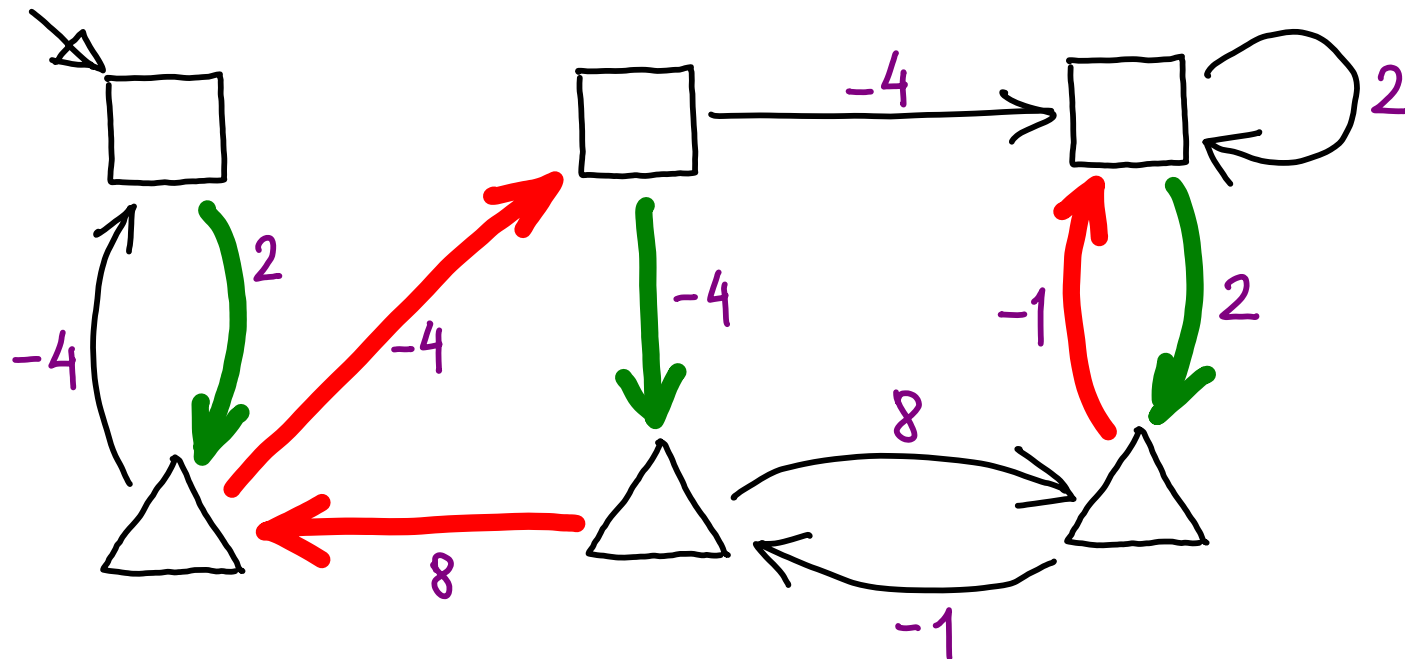
# DISCOUNTED GAMES

$$0 < \delta < 1$$

Players:

MAX =  $\square$

MIN =  $\triangle$



Winner of an infinite play  $\pi = \langle v_0, v_1, v_2, \dots \rangle$

MAX if 
$$\sum_{i=0}^{\infty} \delta^i \cdot r(v_i, v_{i+1}) \geq 0$$

# PAYOFF, VALUE, DETERMINACY

$$\pi = \langle s_0, s_1, s_2, \dots \rangle$$

Mean payoff :  $A(\pi) = \lim_{n \rightarrow \infty} \frac{1}{n} \cdot \sum_{i=0}^{n-1} r(s_i, s_{i+1})$

Discounted total :  $D_\delta(\pi) = \sum_{i=0}^{\infty} \delta^i \cdot r(s_i, s_{i+1})$

---

Lower value :  $Val_*(s) = \sup_{\chi \in \Sigma_{\text{Max}}} \inf_{\mu \in \Sigma_{\text{Min}}} P(\text{Play}(s, \mu, \chi))$

Upper value :  $Val^*(s) = \inf_{\mu \in \Sigma_{\text{Min}}} \sup_{\chi \in \Sigma_{\text{Max}}} P(\text{Play}(s, \mu, \chi))$

---

Determinacy :  $Val(s) = Val_*(s) = Val^*(s)$

# COMPARING COMPUTATIONAL COMPLEXITY OF INFINITE GAMES

THM [PURI 1995]

There is a polynomial-time reduction  
from parity games to mean-payoff games

THM [1960's]

There is a polynomial-time reduction  
from mean-payoff games to discounted games

---

THM [HARDY LITTLEWOOD 1930's]

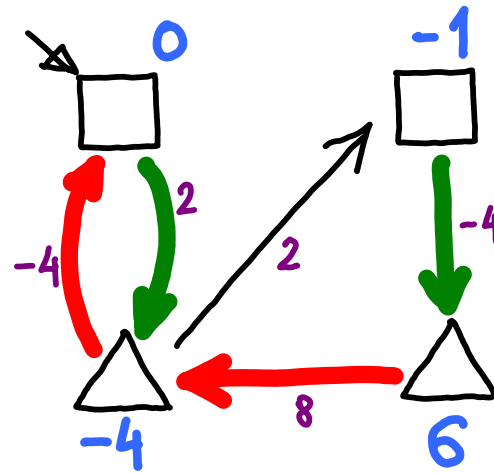
If  $\lim_{n \rightarrow \infty} \left[ \frac{1}{n} \cdot \sum_{i=0}^{n-1} a_i \right]$  exists

then  $\lim_{n \rightarrow \infty} \left[ \frac{1}{n} \cdot \sum_{i=0}^{n-1} a_i \right] = \lim_{\delta \nearrow 1} \left[ (1-\delta) \cdot \sum_{i=0}^{\infty} \delta^i \cdot a_i \right]$

# OPTIMALITY EQUATIONS FOR DISCOUNTED GAMES

OPT(G):

$$v_s = \begin{cases} \max_{(s,t) \in E} (r_{(s,t)} + \delta \cdot v_t) & \text{if } s \in S_{\text{Max}} \\ \min_{(s,t) \in E} (r_{(s,t)} + \delta \cdot v_t) & \text{if } s \in S_{\text{Min}} \end{cases}$$



$$\delta = \frac{1}{2}$$

**LEMMA** If  $V = \text{OPT}(G)$  then  $V = \text{Val}^G$   
and **positional strategies** choosing optimal successor are **optimal**

# OPTIMALITY EQUATIONS FOR DISCOUNTED GAMES

$$F: v_s \mapsto \begin{cases} \max_{(s,t) \in E} \left( r_{(s,t)} + \delta \cdot v_t \right) & \text{if } s \in S_{\text{Max}} \\ \min_{(s,t) \in E} \left( r_{(s,t)} + \delta \cdot v_t \right) & \text{if } s \in S_{\text{Min}} \end{cases}$$

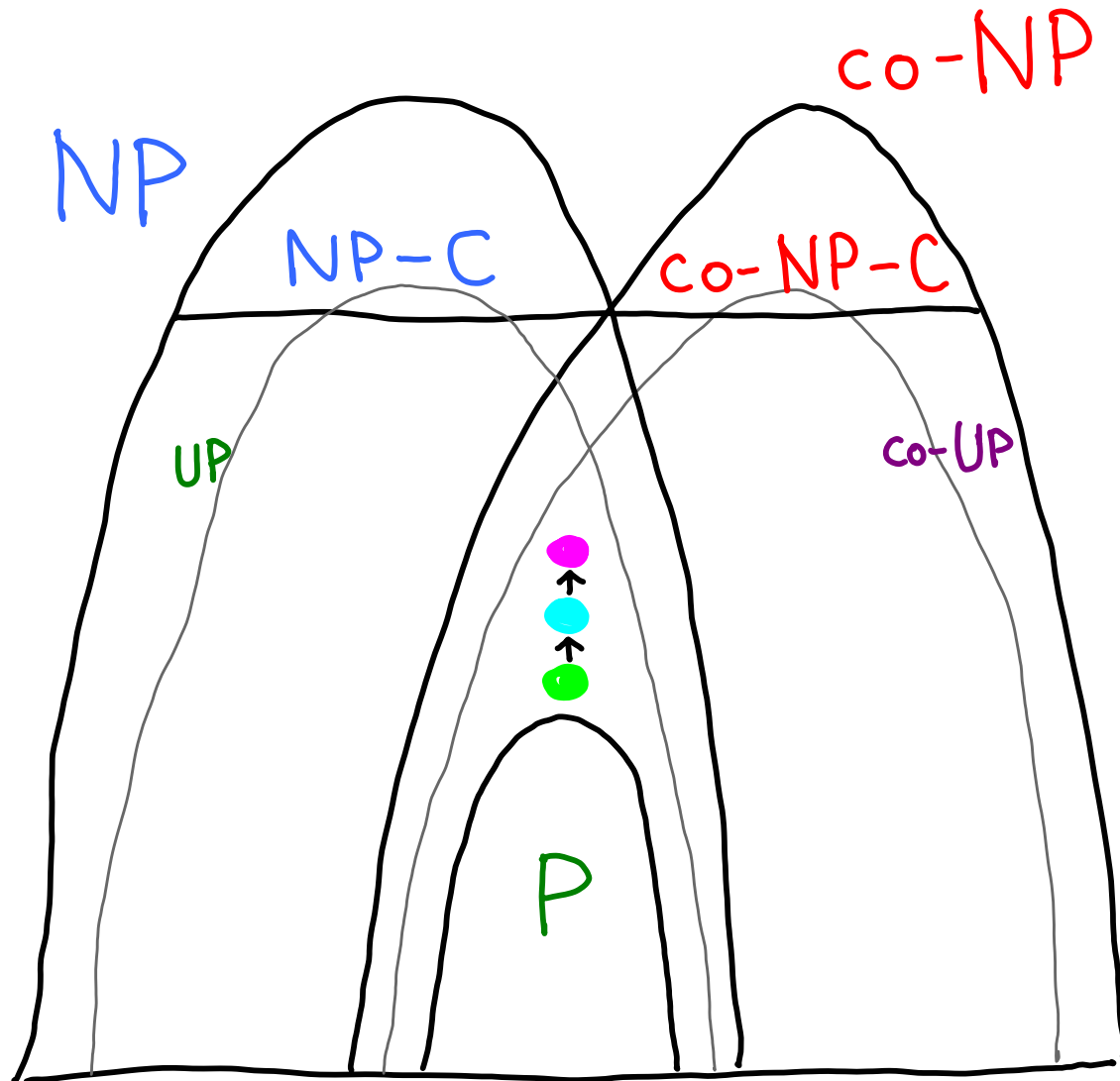
FACT  $F: \mathbb{R}^n \rightarrow \mathbb{R}^n$  is a contraction

## COROLLARY

- $F$  has a **unique fixed point**, i.e.,  $\text{OPT}(G)$  has a solution
- Discounted, mean-payoff, and parity games:
  - are **positionally determined**
  - are in **UP  $\cap$  co-UP**

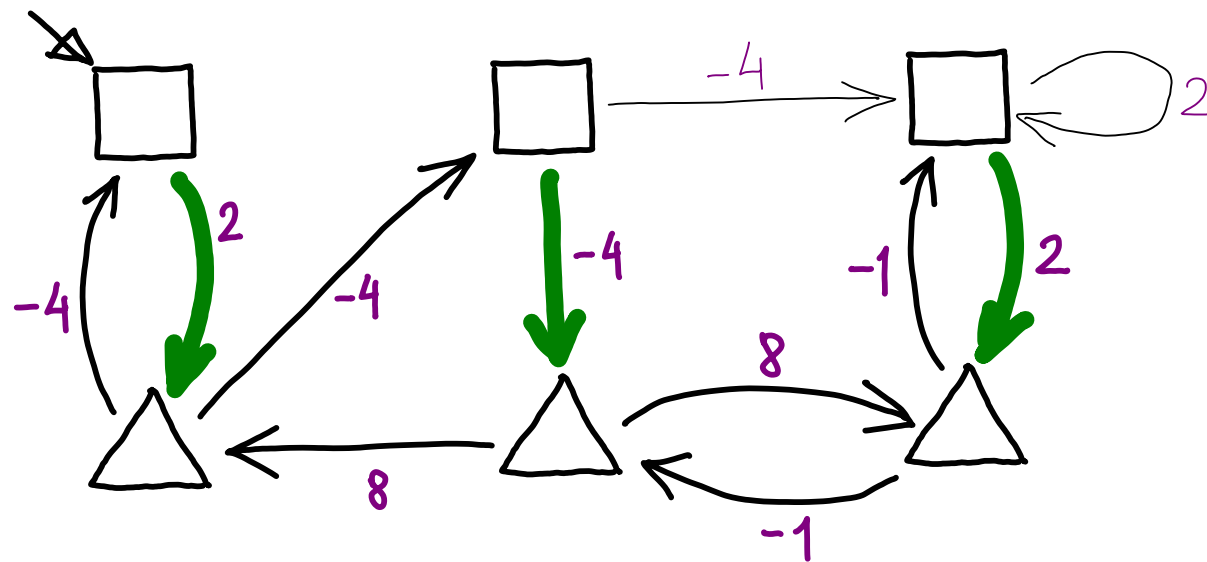


# COMPUTATIONAL COMPLEXITY OF SOLVING INFINITE GAMES



DISCOUNTED GAMES  
AVERAGE-REWARD GAMES  
PARITY GAMES

# 2-PLAYER STRATEGY IMPROVEMENT



$$\delta = \frac{1}{2}$$

0. Pick  $x \in \Pi_{\max}$

1. Find  $V: S \rightarrow \mathbb{R}$ , such that  $V \models \text{OPT}(\Gamma^x)$

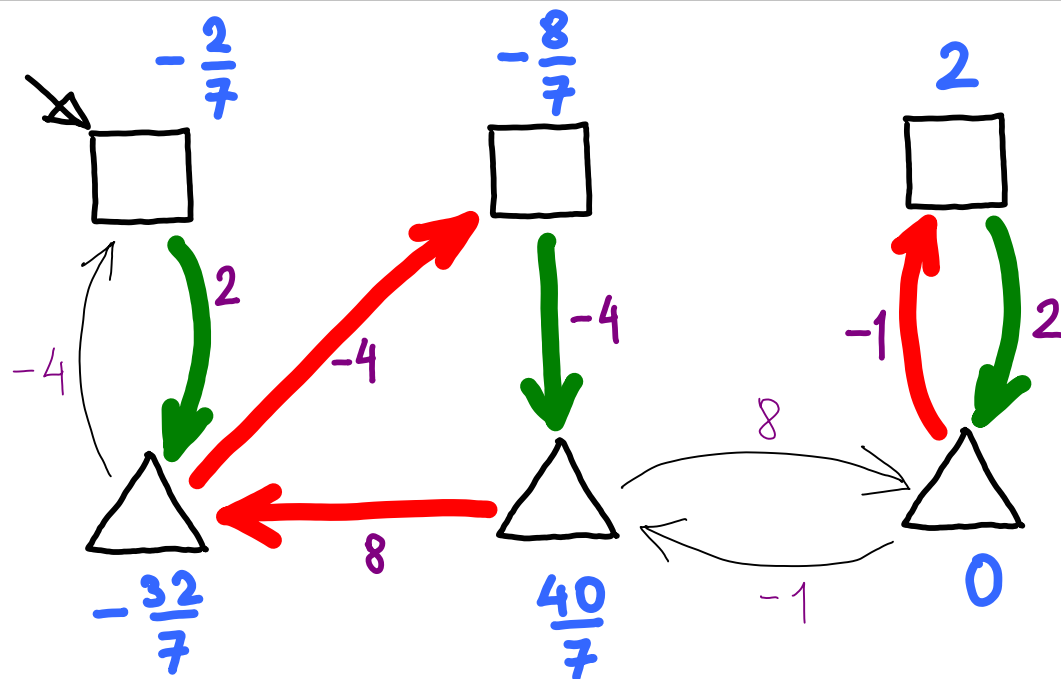
2. If  $V \not\models \text{OPT}(\Gamma)$

then  $x := \text{Improve}(x, V)$ ; goto 1.

Compute the best response  $\mu$  to  $x$  for Min

Improve locally w.r.t. the best response

# 2-PLAYER STRATEGY IMPROVEMENT



$$\delta = \frac{1}{2}$$

0. Pick  $x \in \Pi_{\text{Max}}$

1. Find  $V: S \rightarrow \mathbb{R}$ , such that  $V = \text{OPT}(G^x)$

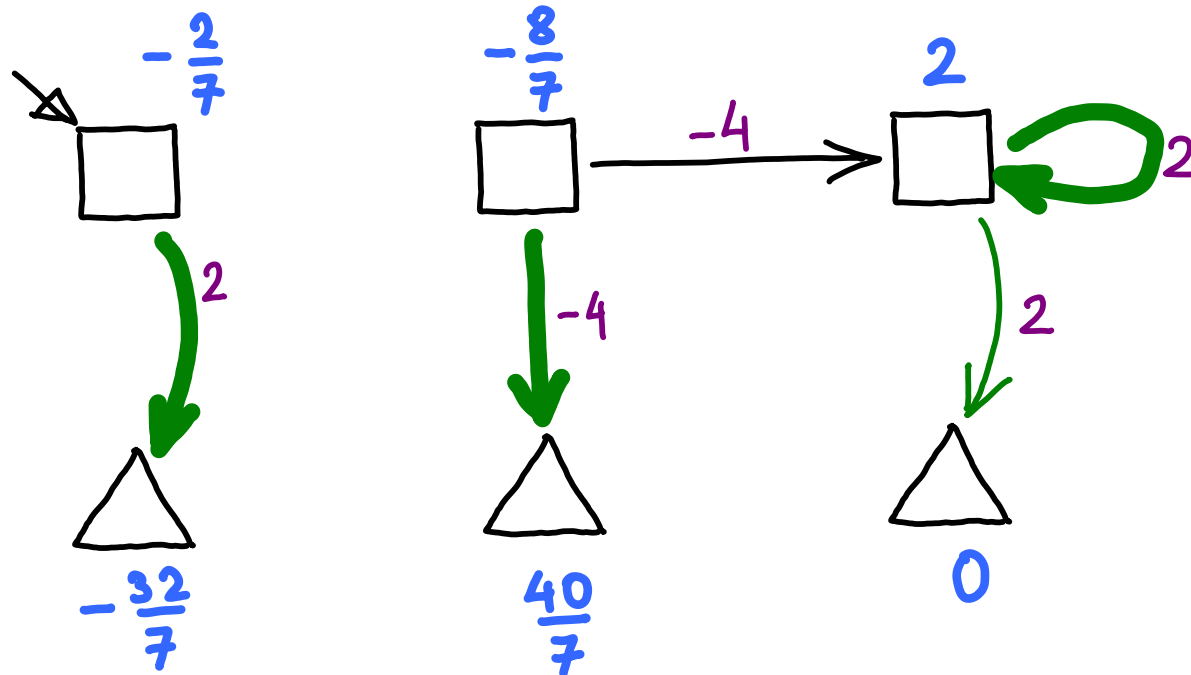
2. If  $V \neq \text{OPT}(G)$

then  $x := \text{Improve}(x, V)$ ; goto 1.

Compute the **best response**  $\mu$  to  $x$  for Min

Improve locally w.r.t. the **best response**

# 2-PLAYER STRATEGY IMPROVEMENT



$$\delta = \frac{1}{2}$$

0. Pick  $x \in \Pi_{\text{Max}}$

1. Find  $V: S \rightarrow \mathbb{R}$ , such that  $V \models \text{OPT}(G^x)$

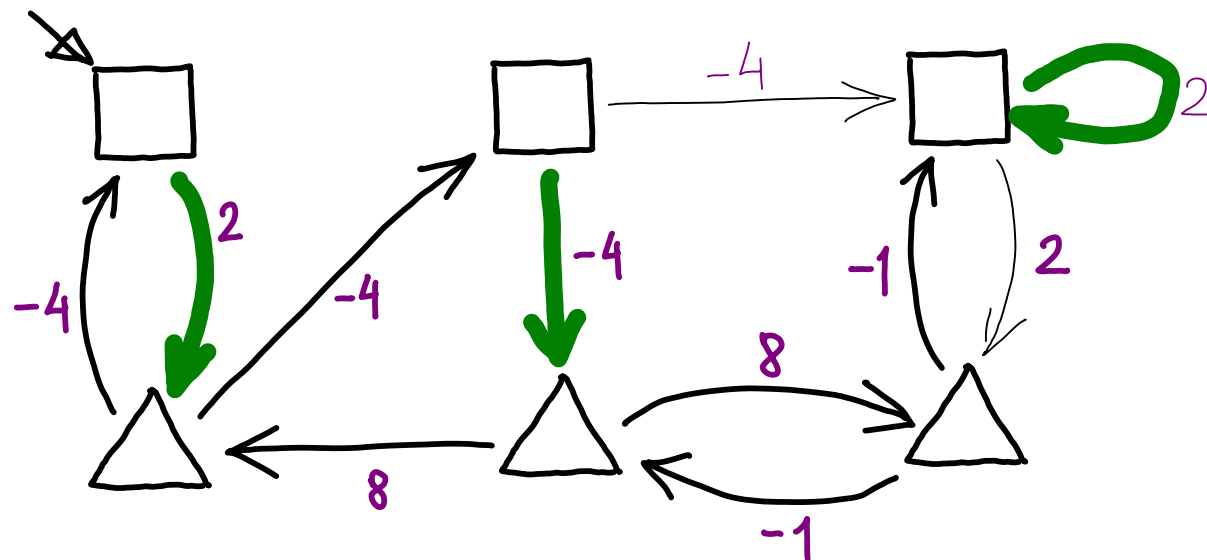
2. **If**  $V \neq \text{OPT}(G)$

**then**  $x := \text{Improve}(x, V)$ ; **goto** 1.

Compute the **best response**  $\mu$  to  $x$  for Min

**Improve** locally w.r.t. the **best response**

# 2-PLAYER STRATEGY IMPROVEMENT



$$\delta = \frac{1}{2}$$

0. Pick  $x \in \Pi_{\text{Max}}$

1. Find  $V: S \rightarrow \mathbb{R}$ , such that  $V = \text{OPT}(G^x)$

2. If  $V \neq \text{OPT}(G)$

then  $x := \text{Improve}(x, V)$ ; goto 1.

Compute the **best response**  $\mu$  to  $x$  for Min

Improve locally w.r.t. the **best response**

# CORRECTNESS OF 2-PLAYER STRATEGY IMPROVEMENT

## IMPROVEMENT LEMMA

If  $V \models \text{OPT}(G^x)$ ,  
 $x' = \text{Improve}(x, V)$ , and  
 $V' \models \text{OPT}(G^{x'})$   
then  $V' \geq V$ , and  
 $V' > V$  if  $x' \neq x$ .

## TERMINATION LEMMA

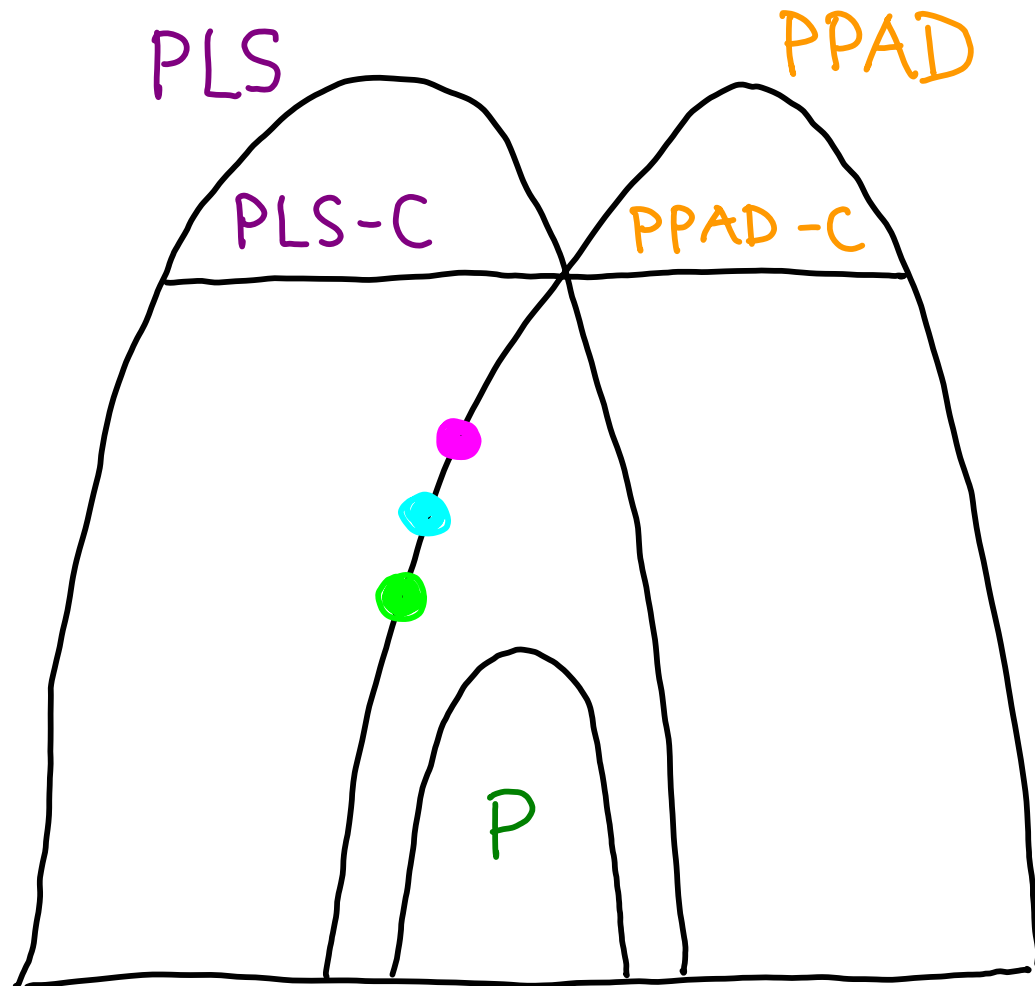
Strategy improvement *terminates* (in  $\leq |\Pi_{\max}|$  steps)  
and returns  $V \models \text{OPT}(G)$ .

---

## COROLLARY

Computing the value of *discounted, average-reward,*  
and *parity* games is in **PLS**

# COMPUTATIONAL COMPLEXITY OF SOLVING INFINITE GAMES



DISCOUNTED GAMES

AVERAGE-REWARD GAMES

PARITY GAMES

# LINEAR COMPLEMENTARITY PROBLEM

Given:  $M \in \mathbb{R}^{n \times n}$

$q \in \mathbb{R}^n$

Find:  $z, w \in \mathbb{R}^n$

such that

linear  $\begin{cases} z \geq 0 \\ w \geq 0 \\ w = Mz + q \end{cases}$

complementarity  $\begin{cases} z \perp w \end{cases}$

## FACT

If  $q \geq 0$

then  $\begin{pmatrix} 0 \\ z \end{pmatrix}, \begin{pmatrix} q \\ w \end{pmatrix} = \text{LCP}(M, q)$

i.e.,  $z^T \cdot w = 0$



# COMPLEMENTARY CONES OF LCP

$$z \geq 0 \quad \perp \quad w = Mz + q \geq 0$$

iff

$$z \geq 0 \quad \perp \quad w \geq 0 \quad \text{and} \quad q = -Mz + Iw$$

iff

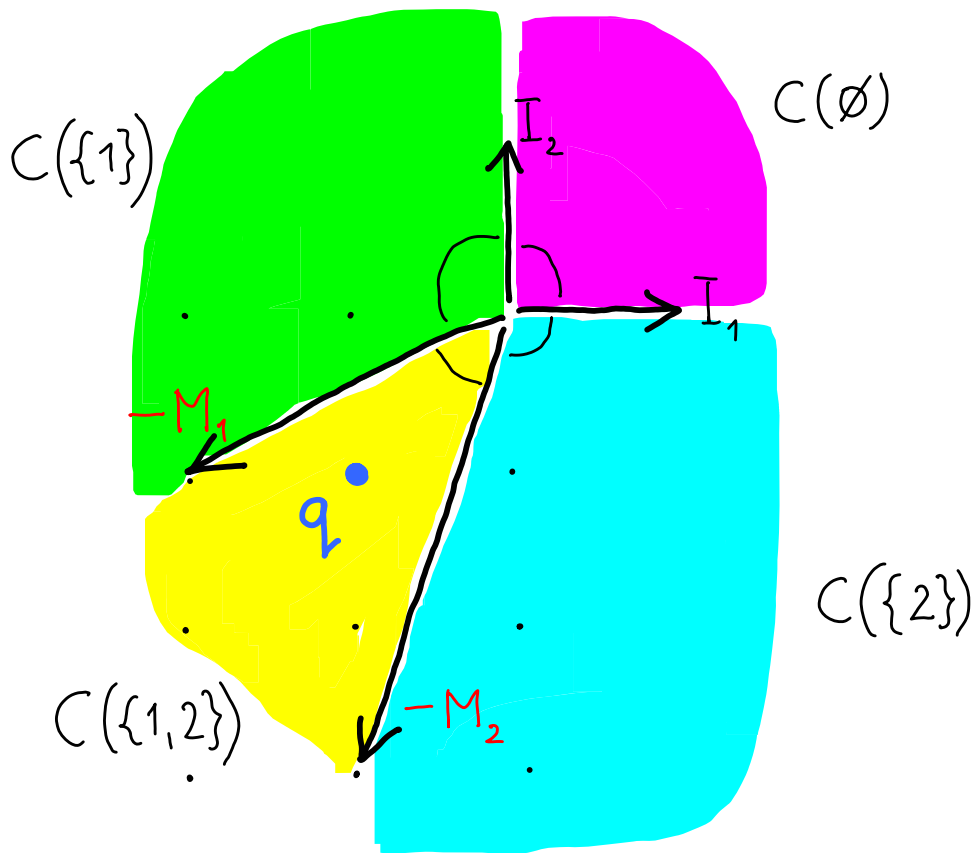
$$q \in C(\alpha) = \text{cone} \left( \{-M_k : k \in \alpha\} \cup \{I_k : k \notin \alpha\} \right)$$

for some  $\alpha = \{k : w_k = 0\} \subseteq \{1, \dots, n\}$

# COMPLEMENTARY CONES OF A MATRIX

$$z \geq 0 \perp w \geq 0 \quad \text{and} \quad q = -Mz + Iw$$

$$M = \begin{pmatrix} M_1 & M_2 \\ 2 & 1 \\ 1 & 3 \end{pmatrix} \quad q = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$$



$$q \in C(\{1,2\})$$

$$\Downarrow$$

$$(z_1, z_2, 0, 0) \in \text{LCP}(M, q)$$

where  $q = -Mz$

# P-MATRICES

DEF  $M \in \mathbb{R}^{n \times n}$  is a P-matrix  
if all its principal minors are positive

THM  $M$  is a P-matrix  
iff  $\text{LCP}(M, q)$  has a unique solution for every  $q \in \mathbb{R}^n$ .

---

## EXAMPLE

$$M = \begin{pmatrix} 2 & 1 \\ 1 & 3 \end{pmatrix}$$

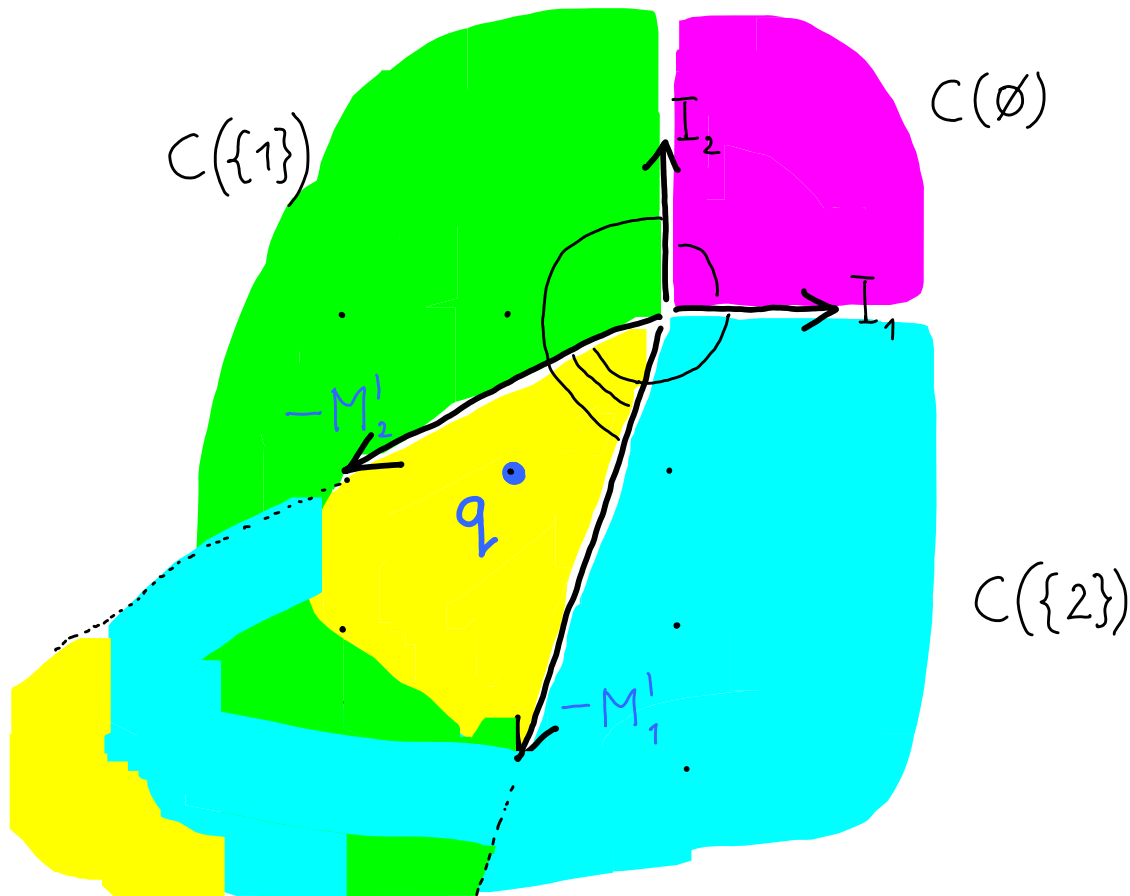
$$M' = \begin{pmatrix} 1 & 2 \\ 3 & 1 \end{pmatrix}$$

- $M$  is a P-matrix :  
 $\det(M_{11}) = 2$   
 $\det(M_{22}) = 3$   
 $\det(M) = 5$
- $M'$  is not a P-matrix :  $\det(M') = -5$

# COMPLEMENTARY CONES OF A NON P-MATRIX

$$z \geq 0 \perp w \geq 0 \quad \text{and} \quad q = -Mz + Iw$$

$$M' = \begin{pmatrix} \overset{M'_1}{1} & \overset{M'_2}{2} \\ 3 & 1 \end{pmatrix} \quad q = \begin{pmatrix} -1 \\ -1 \end{pmatrix}$$



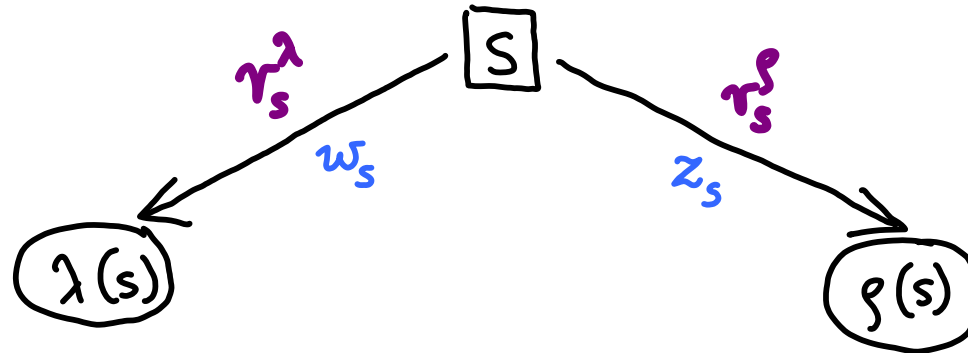
$$q \in C(\{1,2\})$$

$$q \in C(\{1\})$$

$$q \in C(\{2\})$$

# OPTIMALITY EQUATIONS AND COMPLEMENTARITY

$$v_s = \max \left\{ r_s^\lambda + \delta \cdot v_{\lambda(s)}, r_s^\rho + \delta \cdot v_{\rho(s)} \right\}$$



slacks

$$v_s = w_s + r_s^\lambda + \delta \cdot v_{\lambda(s)}$$

$$v_s = z_s + r_s^\rho + \delta \cdot v_{\rho(s)}$$

$$z_s, w_s \geq 0$$

$$z_s \cdot w_s = 0$$

complementarity

# OPTIMALITY EQUATIONS AND COMPLEMENTARITY

$$s \in S_{\max}: \quad v_s = \max \left\{ r_s^\lambda + \delta \cdot v_{\lambda(s)}, r_s^\rho + \delta \cdot v_{\rho(s)} \right\}$$

$$s \in S_{\min}: \quad v_s = \min \left\{ r_s^\lambda + \delta \cdot v_{\lambda(s)}, r_s^\rho + \delta \cdot v_{\rho(s)} \right\}$$

Replace max/min with *slacks* and *complementarity*

$s \in S_{\max}:$

$$v_s = w_s + r_s^\lambda + \delta \cdot v_{\lambda(s)}$$

$$v_s = z_s + r_s^\rho + \delta \cdot v_{\rho(s)}$$

*slacks*

$s \in S_{\min}:$

$$v_s = -w_s + r_s^\lambda + \delta \cdot v_{\lambda(s)}$$

$$v_s = -z_s + r_s^\rho + \delta \cdot v_{\rho(s)}$$

$s \in S:$

$$w_s \geq 0 \quad \perp \quad z_s \geq 0$$

*complementarity*

# REDUCTION TO LCP: REWRITE

$s \in S_{\max}$ :

$$v_s = w_s + r_s^\lambda + \delta \cdot v_{\lambda(s)}$$

$$v_s = z_s + r_s^\rho + \delta \cdot v_{\rho(s)}$$

stacks

$s \in S_{\min}$ :

$$v_s = -w_s + r_s^\lambda + \delta \cdot v_{\lambda(s)}$$

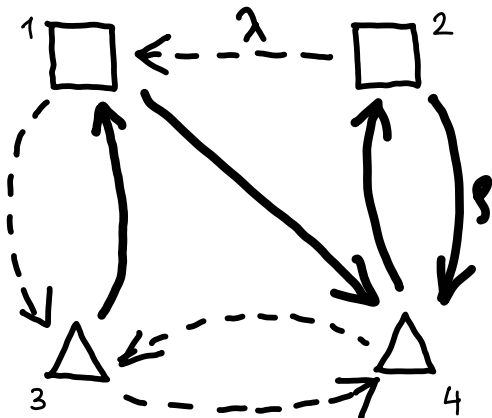
$$v_s = -z_s + r_s^\rho + \delta \cdot v_{\rho(s)}$$



$$(\hat{I} - \delta \cdot \hat{T}^\lambda) v = w + \hat{I} r^\lambda$$

$$(\hat{I} - \delta \cdot \hat{T}^\rho) v = z + \hat{I} r^\rho$$

## EXAMPLE



$$\hat{I} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \end{pmatrix}$$

$$\hat{T}^\rho = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ -1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \end{pmatrix}$$

# REDUCTION TO LCP: ELIMINATE $v$

$$(\hat{I} - \delta \cdot \hat{T}^\lambda) v = w + \hat{I} r^\lambda$$

$$(\hat{I} - \delta \cdot \hat{T}^s) v = z + \hat{I} r^s$$

Eliminate  $v$ :  $w + \hat{I} r^\lambda = \underbrace{(\hat{I} - \delta \cdot \hat{T}^\lambda) \cdot (\hat{I} - \delta \cdot \hat{T}^s)^{-1}}_M \cdot (z + \hat{I} r^s)$

$$w = Mz + q$$

$$w \geq 0 \quad \perp \quad z \geq 0$$

$$M = (\hat{I} - \delta \cdot \hat{T}^\lambda) \cdot (\hat{I} - \delta \cdot \hat{T}^s)^{-1}$$

$$q = M(\hat{I} r^s) - (\hat{I} r^\lambda)$$



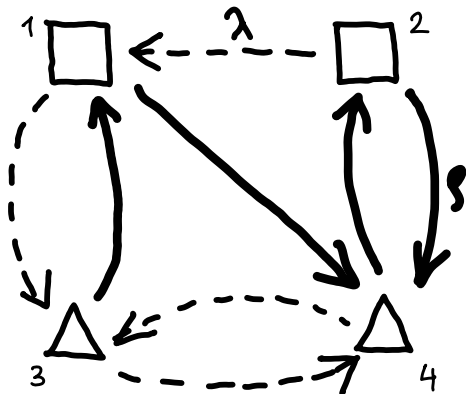
# REDUCTION TO LCP: ELIMINATE $v$

$$w = Mz + q$$
$$w \geq 0 \quad \perp \quad z \geq 0$$

$$M = (\hat{I} - \delta \cdot \hat{T}^\lambda) \cdot (\hat{I} - \delta \cdot \hat{T}^\beta)^{-1}$$

$$q = M(\hat{I} r^\beta) - (\hat{I} r^\lambda)$$

## EXAMPLE



$$\hat{I} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & -1 \end{pmatrix}$$

$$\hat{T}^\beta = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ -1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \end{pmatrix}$$

$$(\hat{I} - \delta \cdot \hat{T}^\beta) = \begin{pmatrix} 1 & 0 & 0 & -\delta \\ 0 & 1 & 0 & -\delta \\ \delta & 0 & -1 & 0 \\ 0 & \delta & 0 & -1 \end{pmatrix}$$

# STRICTLY DIAGONALLY DOMINANT MATRICES

FACT [Levy-Desplanques]

If  $A \in \mathbb{R}^{n \times n}$  is strictly diagonally dominant,  
then  $A$  is non-singular.

FACT  $(\hat{I} - \delta \cdot \hat{T}^\lambda)$  and  $(\hat{I} - \delta \cdot \hat{T}^\beta)$  are s.d.d.

COROLLARY  $M = (\hat{I} - \delta \hat{T}^\lambda) \cdot (\hat{I} - \delta \hat{T}^\beta)^{-1}$  is well-defined

# P-MATRICES OF THE FORM $B \cdot C^{-1}$

THM [Johnson - Tsatsomeros '95]

Let  $M = B \cdot C^{-1}$ , where  $B, C \in \mathbb{R}^{n \times n}$

Then,  $M$  is a P-matrix

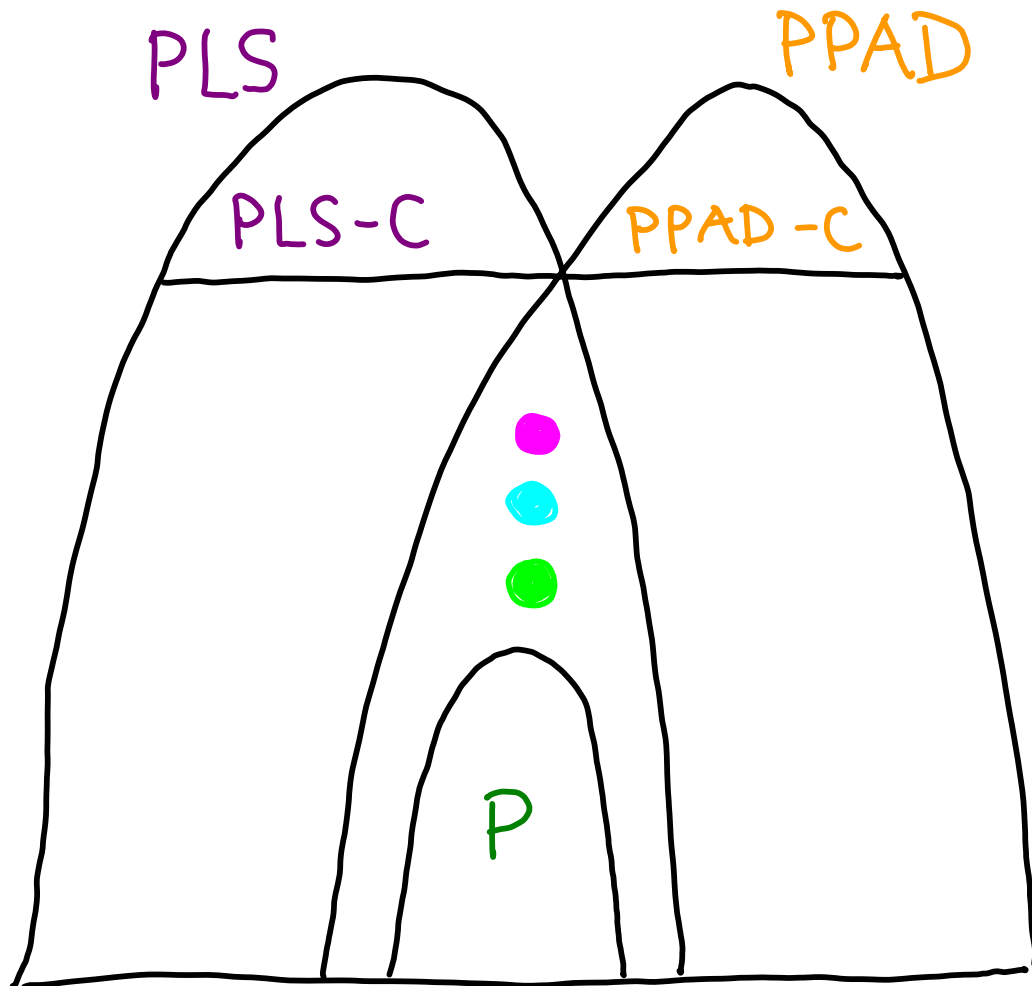
if  $T \cdot B + (I - T) \cdot C$  is non-singular for all  $T \in [0, I]$

## COROLLARY

•  $M = (\hat{I} - \delta \cdot \hat{T}^\lambda) \cdot (\hat{I} - \delta \cdot \hat{T}^\rho)^{-1}$  is a P-matrix.

• Computing the value of discounted, average-reward, and parity games is in PPAD

# COMPUTATIONAL COMPLEXITY OF SOLVING INFINITE GAMES



DISCOUNTED GAMES  
AVERAGE-REWARD GAMES  
PARITY GAMES

# WHAT IS $q$ ?

## LEMMA

Let  $v^s$  be the vector of values of strategy pair  $s$ , i.e.,

$$v^s = r^s + \delta \cdot T^s \cdot v^s$$

Then:

$$q = \hat{I} \left( (r^s + \delta \cdot T^s \cdot v^s) - (r^\lambda + \delta \cdot T^\lambda \cdot v^s) \right)$$

take "right" edge  
and then follow  $s$

take "left" edge  
and then follow  $s$

# ALGORITHM 1: STRATEGY IMPROVEMENT REVISITED

DEF.  $s \in S_{\text{Max}}$  is *switchable* (for strategy pair  $\rho$ ) if

$$v_s^\rho + \delta \cdot v_{\rho(s)}^\rho < v_s^\lambda + \delta \cdot v_{\lambda(s)}^\rho$$

1. Start with arbitrary  $\rho, \lambda$ .
2. While there is a *switchable* vertex (for  $\rho$ ) do
3. Find  $\rho'$  such that:
  - a)  $\rho' \upharpoonright S_{\text{Max}} = \rho \upharpoonright S_{\text{Max}}$
  - b) no  $s \in S_{\text{Min}}$  is *switchable* for  $\rho'$

}  $\rho' \upharpoonright S_{\text{Min}}$  is a best response to  $\rho \upharpoonright S_{\text{Max}}$
4. Switch all (or some) switchable vertices in  $S_{\text{Max}}$

# ALGORITHM 2: MURTY'S "LEAST-INDEX" METHOD

1. Fix a permutation ("indexing") of states.
  2. While there is a **switchable** vertex do
  3.     Switch the **switchable** vertex with **least index**.
- 

THM Murty's algorithm terminates (in  $\leq 2^n$  steps).

---

"Nested" strategy improvement:  $^1\triangle$   $^2\triangle$   $^3\triangle$   $^4\square$   $^5\square$

New algorithm:  $^1\square$   $^2\triangle$   $^3\triangle$   $^4\square$   $^5\triangle$

# ALGORITHM 3: COTTLE-DANZIG

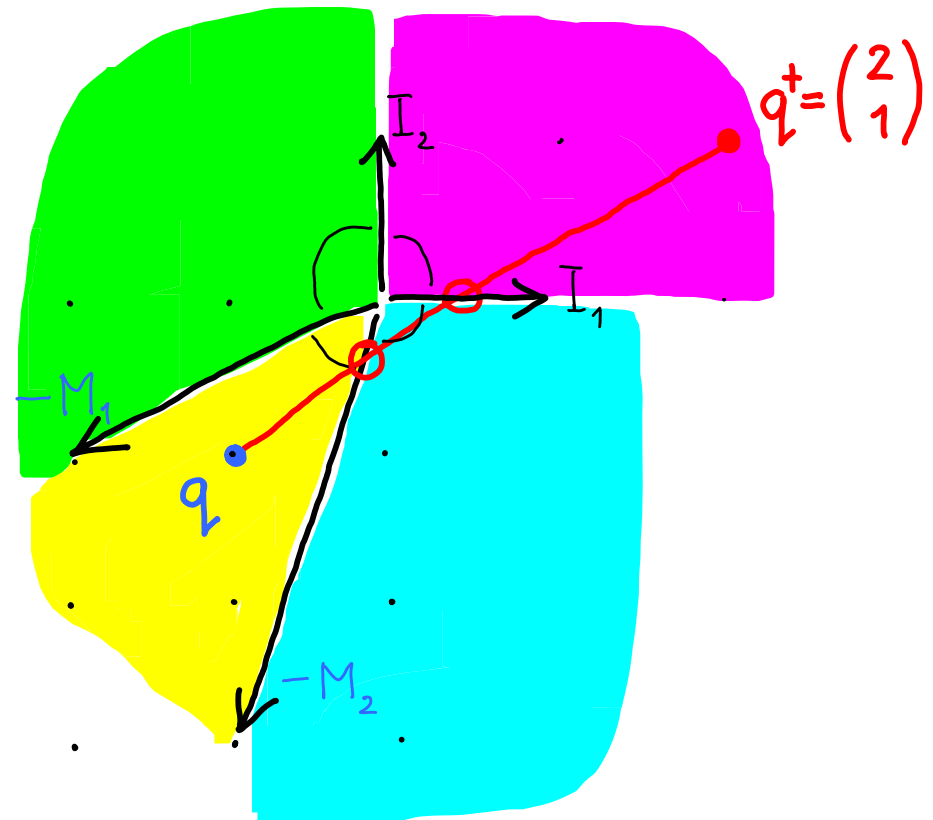
1. For  $s = 1, 2, \dots, n$  do
2. If  $s$  is **switchable** then
3.     Drive  $r_s^p$  until  $s$  becomes indifferent
4.     While driving  $r_s^p$ ,  
       switch a vertex in  $\{1, 2, \dots, s-1\}$   
       when it becomes indifferent
5.     Switch  $s$  and restore  $r_s^p$

**FACT** Cottle-Danzig terminates (in  $\leq 2^m$  steps)



# ALGORITHM 4 : LEMKE

1. Drive  $r^\lambda$  ("linearly"), until no vertex is switchable.
  2. Drive  $r^\lambda$  ("linearly") back, until original  $r^\lambda$  is restored.
- While driving  $r^\lambda$  back,  
switch vertices when they become indifferent.



$$q = M(\hat{I}_{r^p}) - (\hat{I}_{r^\lambda})$$

# EXPONENTIAL TIME LOWER BOUNDS

THM [SAVANI, VON STENGEL 2004]

Lemke-Howson algorithm needs exponential time to find a Nash equilibrium

THM [FEARNLEY, J., SAVANI 2009]

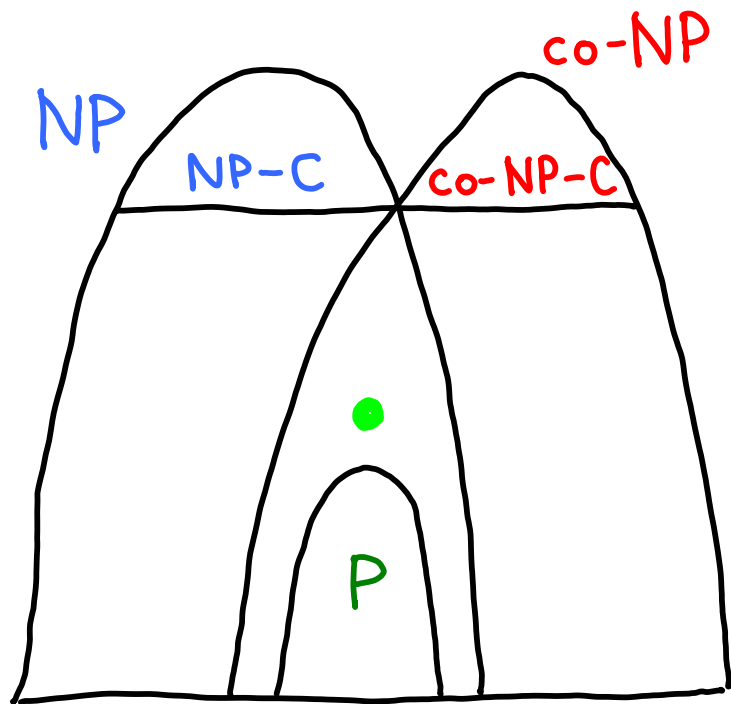
Murty's, Cottle-Danzig, and Lemke's algorithms need exponential time to solve parity games

# LEMKE CAN BE FAST?

THM [ADLER, MEGGIDO 1985]

There is an implementation of Lemke's algorithm that terminates in quadratic number of steps on random linear programs

# How DIFFICULT IS IT TO FIND A WINNING STRATEGY?



P-MATRIX LCP

↑  
COMPUTING  
OPTIMAL  
STRATEGIES

