

# Stackelberg-Pareto Synthesis

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Joint work with

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November 17  
Journées GT Vérif

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1. Reactive Synthesis
2. Stackelberg-Pareto Synthesis
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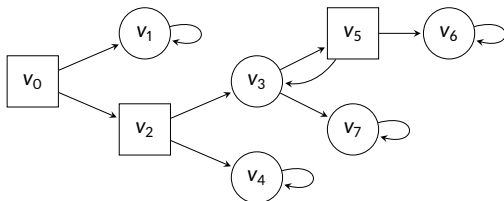
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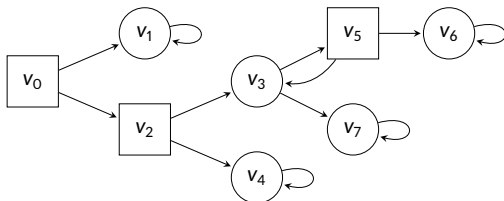
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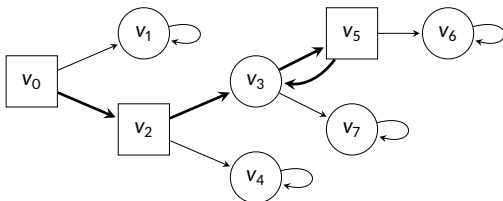
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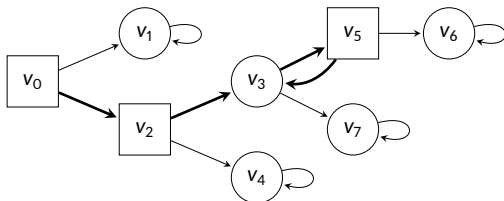
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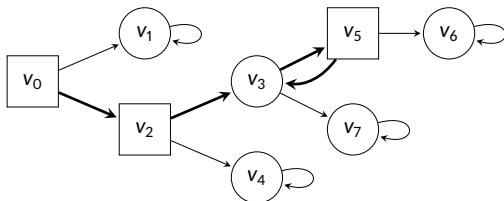
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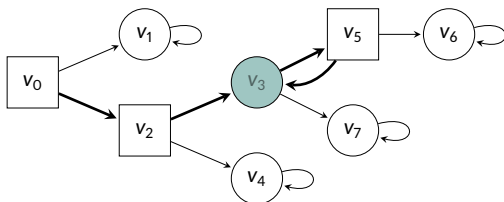
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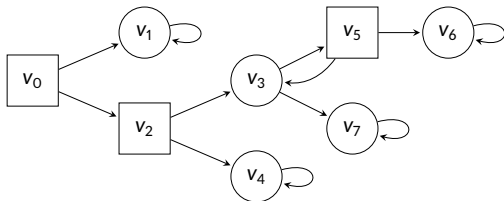
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- **reachability:** plays which visit  $T \subseteq V$

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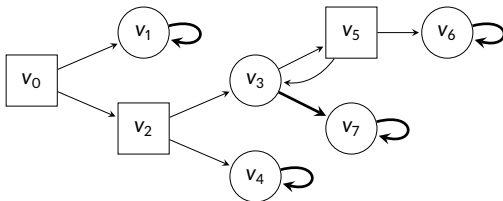
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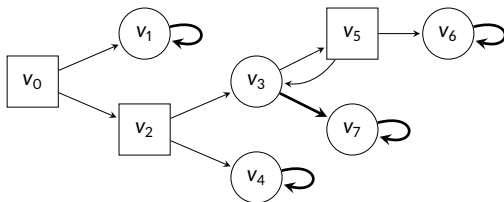
→ given  $\underbrace{v_0 v_1 \dots}_{h} \underbrace{v_k}_{\in V_i}$  yields  $v_{k+1}$  from  $h v_k$  (memory) or  $v_k$  (without)



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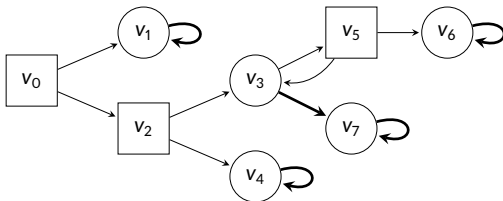


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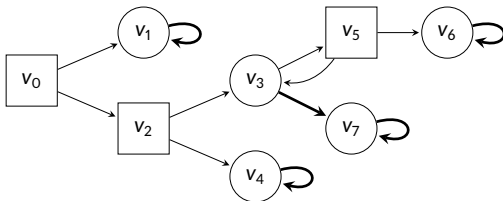
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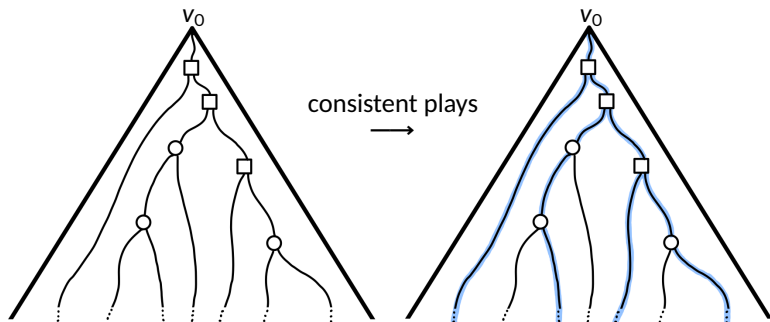
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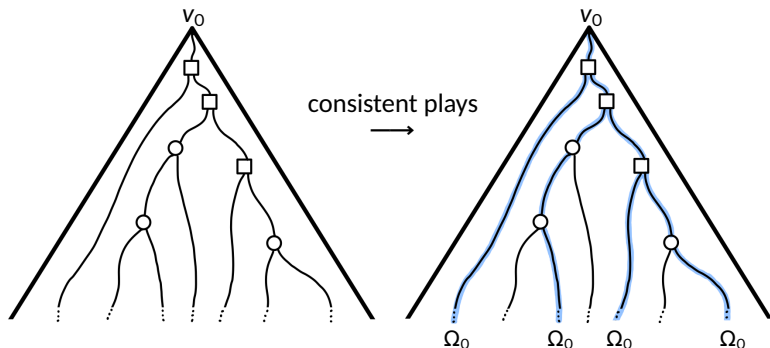
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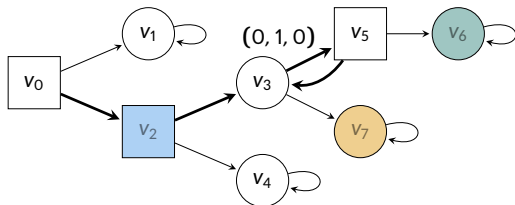
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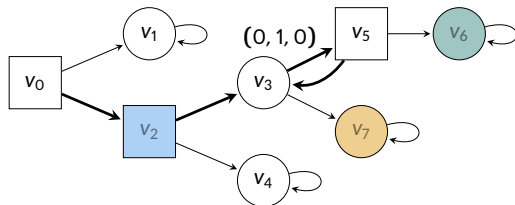
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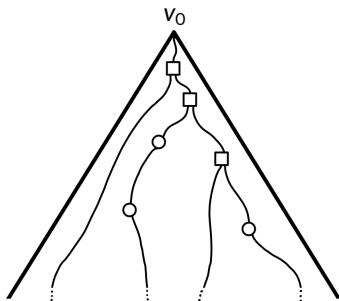


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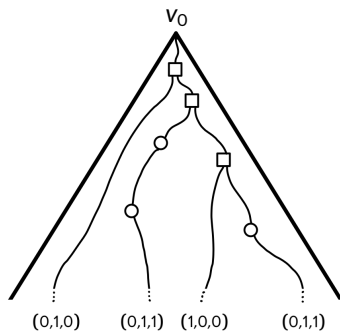
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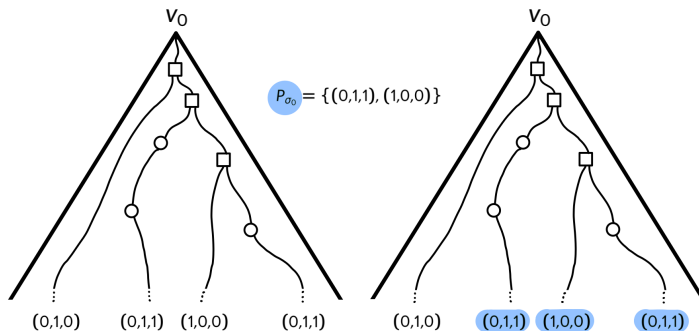
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  - identify **Pareto-optimal** (PO) payoffs (maximal w.r.t.  $\leq$ ) : set  $P_{\sigma_0}$



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The SPS problem is to decide whether there exists a strategy  $\sigma_0$  for Player 0 such that for every play  $\rho \in \text{Plays}_{\sigma_0}$  with  $\text{pay}(\rho) \in P_{\sigma_0}$ , it holds that  $\rho \in \Omega_0$

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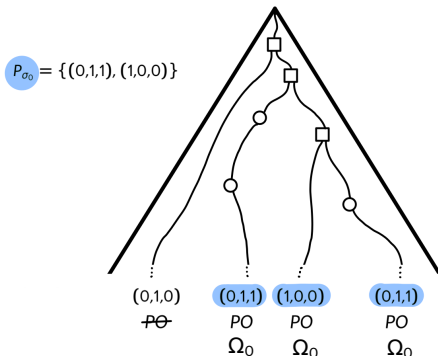
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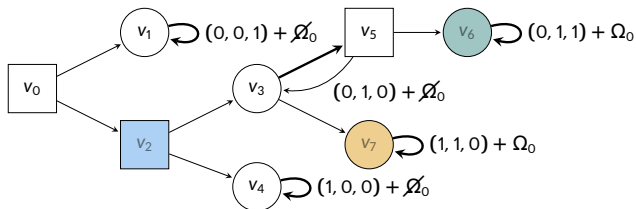
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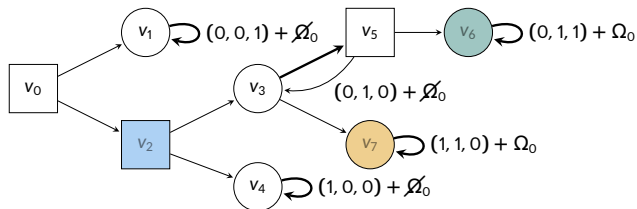
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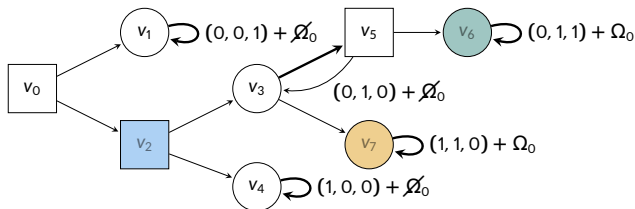
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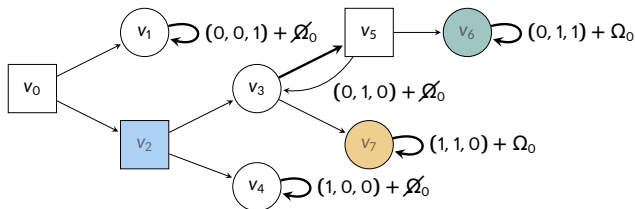


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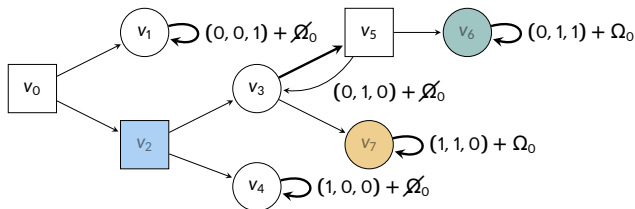


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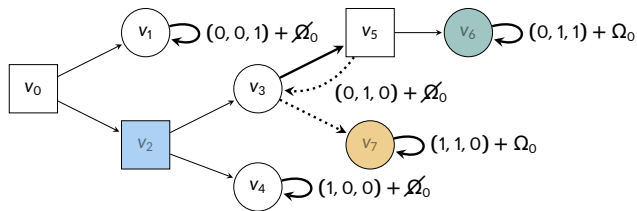


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 $\rightarrow$  the only other **memoryless** strategy is **not a solution either**

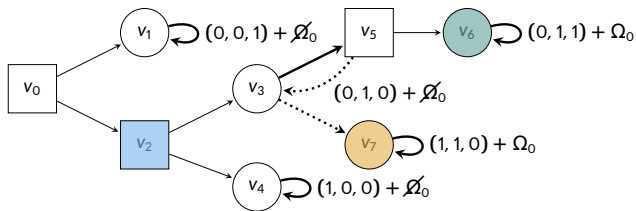
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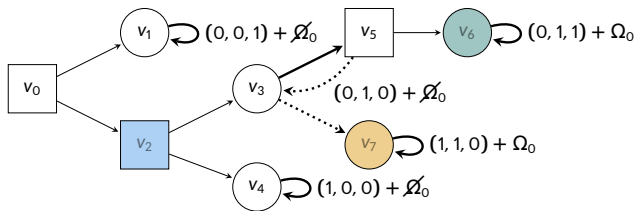
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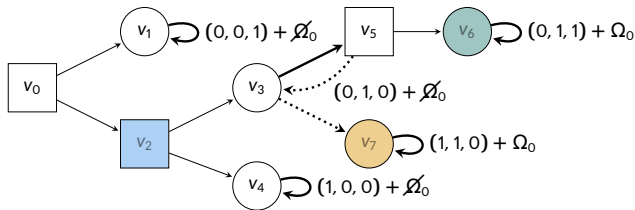
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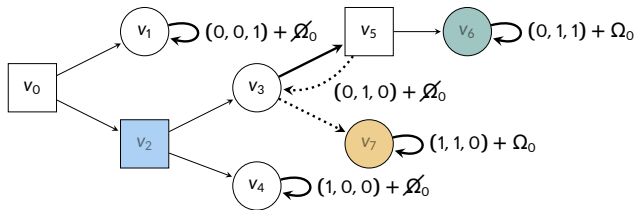


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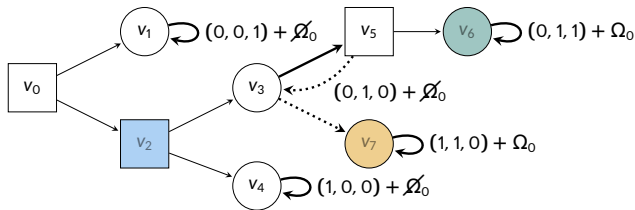
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- parity (**parity SP games**): models general class of  $\omega$ -regular objectives
- reachability (**reachability SP games**): simpler setting



# Outline

1. Reactive Synthesis
2. Stackelberg-Pareto Synthesis
- 3. Our Results**

# Our Results on SP games

## NEXPTIME-Completeness of the SPS problem

The SPS problem is NEXPTIME-complete for reachability SP games and for parity SP games

## Fixed-Parameter Complexity of the SPS problem

Solving the SPS problem is FPT for reachability SP games for parameter  $t$  (number of objectives of Player 1) and FPT for parity SP games for parameters  $t$  and the maximal priority according to each parity objective of Player 1

**Sound:** in practice, we can assume those parameters to have **small values**

NEXPTIME algorithm **not FPT** & FPT algorithm **not usable** for membership

# Complexity Class

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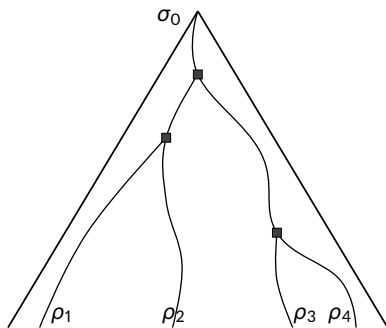
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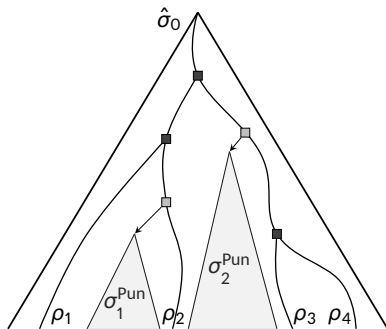
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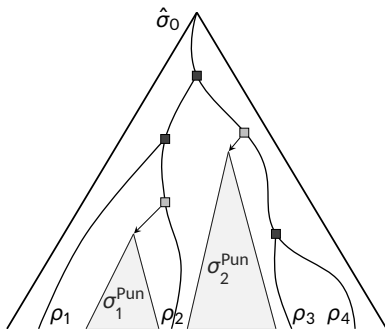


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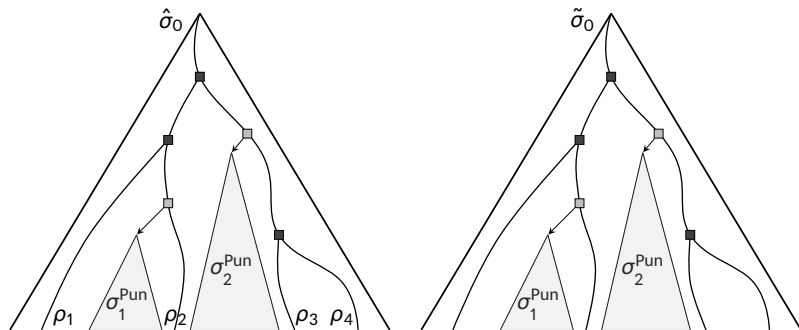
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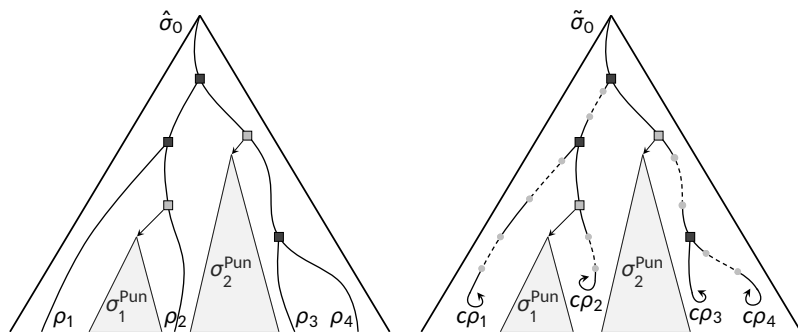
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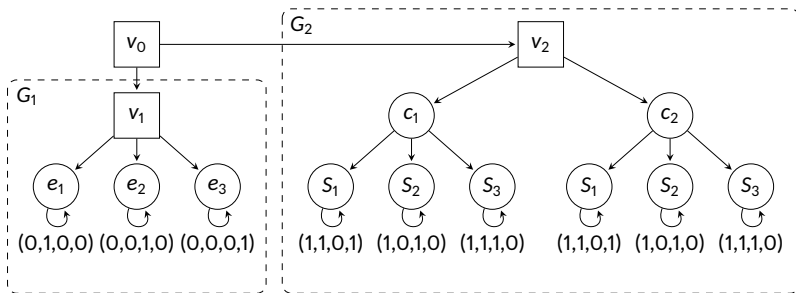
Player 0 has a **solution** to the **SPS problem**  $\Leftrightarrow$  **solution** to the **SC problem**

## Reduction to the Set Cover Problem

$$C = \{e_1, e_2, e_3\}, S_1 = \{e_1, e_3\}, S_2 = \{e_2\}, S_3 = \{e_1, e_2\}, k = 2$$

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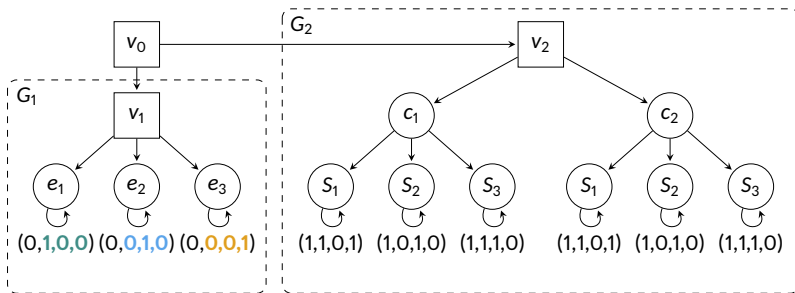
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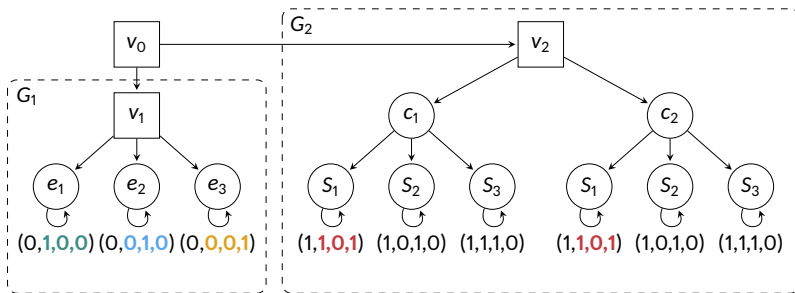
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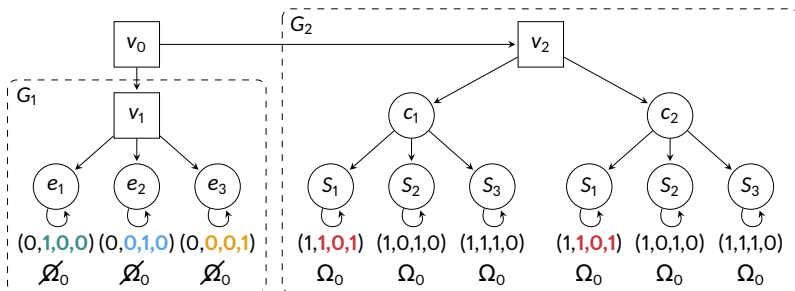
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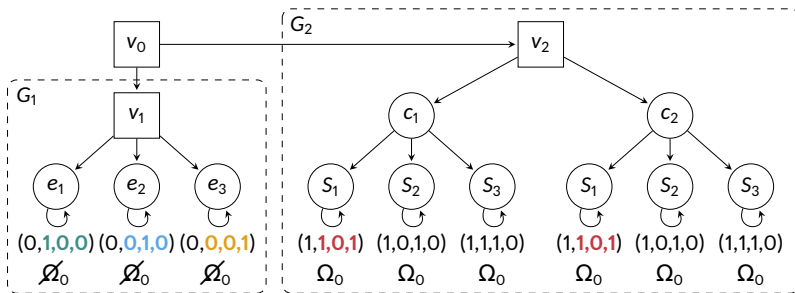
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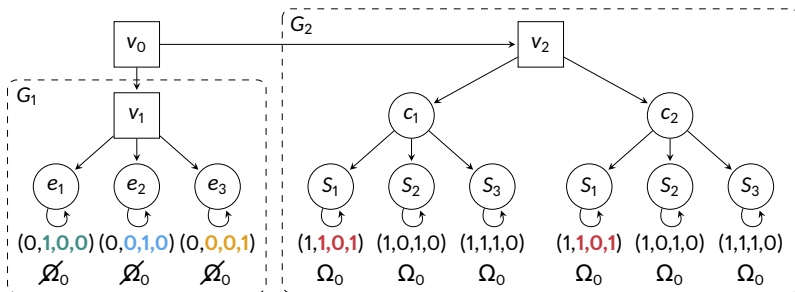
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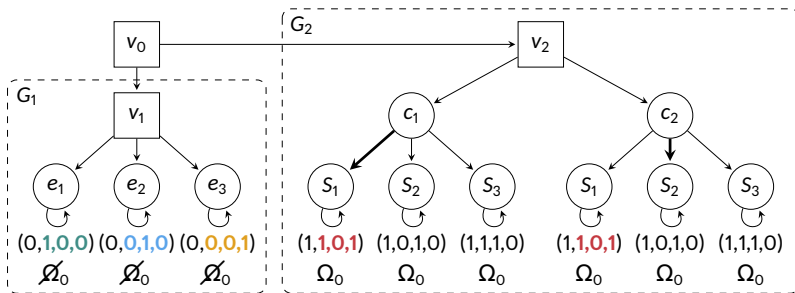
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Each payoff in  $G_1$  must be **< than some payoff in  $G_2$**  (corresponding to a set)

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Intuition: use **succinct variant** of Set Cover problem (NEXPTIME-complete)  
→ Set Cover problem succinctly defined using **CNF formulas**



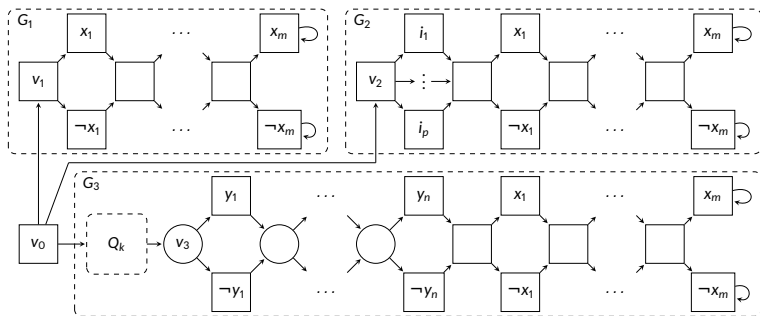
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# Challenger-Prover Game

To show FPT results: reduction to **Challenger-Prover game** (C-P game)

- **two-player zero-sum game**  $\mathcal{G}'$ , created from  $\mathcal{G}$
- played between **Challenger** ( $\mathcal{C}$ ) and **Prover** ( $\mathcal{P}$ )
- **solution** to the SPS problem in  $\mathcal{G} \iff$  **winning strategy** for  $\mathcal{P}$  in  $\mathcal{G}'$
- described in a **generic way**, later **adapted** to parity/reachability

Intuition:  $\mathcal{P}$  tries to **show** the existence of a solution,  $\mathcal{C}$  tries to **disprove** it

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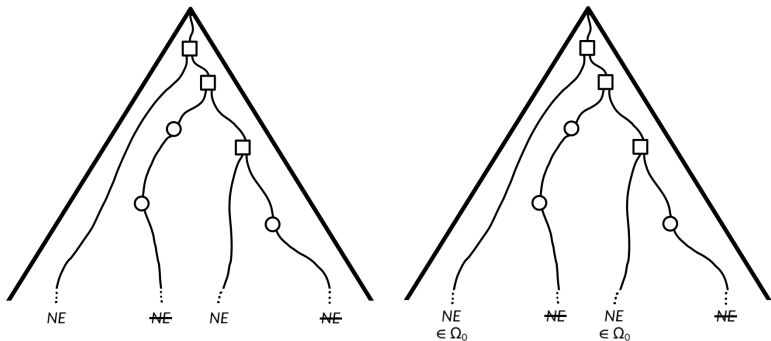
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Wien und Berlin, J. Springer, 1937.

# Adversarial Rational Synthesis [FKL10, KPV16]

- **multiplayer** game
  - Player 0 = system, Players 1 to  $n$  = **components** of environment
  - **rationality**: Players 1 to  $n$  settle to a **Nash Equilibrium** (NE), given  $\sigma_0$
- Player 0 **must satisfy** his objective when the environment plays **any** NE



**Setbacks:** components are independent selfish individuals, no cooperation

## Fixed-Parameter Complexity [DF12]

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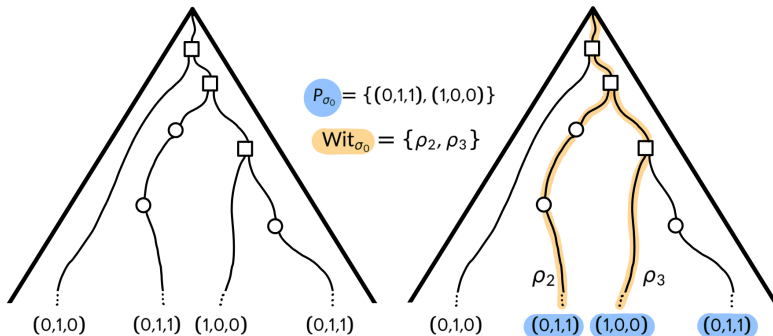
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# Witnesses

C-P game uses important notion of **witness**

- given  $\sigma_0$ , we have the set  $P_{\sigma_0}$  of **PO payoffs**
- for each  $p \in P_{\sigma_0}$ , **there exists**  $\rho$  s.t.  $\text{pay}(\rho) = p$
- select **one such**  $\rho$  for each  $p \in P_{\sigma_0}$  (**witness of  $p$** ): set  $\text{Wit}_{\sigma_0}$

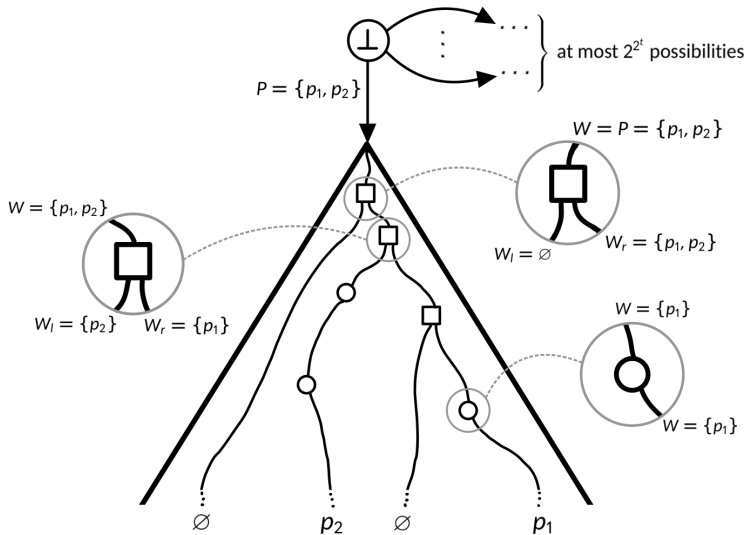


## Intuition on the C-P game

w.l.o.g. we consider SP games s.t. each vertex has at most **two successors**

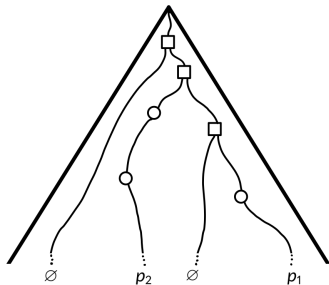
1.  $\mathcal{P}$  **selects a set  $P$**  of payoffs, he announces it is  $P_{\sigma_0}$  for  $\sigma_0$  he is building
2.  $\mathcal{P}$  tries to show the existence of a **set of witnesses for  $P$**
3. After selection, **one-to-one correspondence** between plays in  $\mathcal{G}$  and  $\mathcal{G}'$ 
  - vertices in  $\mathcal{G}'$  are **augmented with a set  $W$**  which is a subset of  $P$
  - initially  $W = P$
  - after history in  $\mathcal{G}'$ ,  $W$  **contains  $p$**  if the **corresponding history** in  $\mathcal{G}$  is **prefix of the witness for  $p$**  in the set  $\text{Wit}_{\sigma_0}$  that  $\mathcal{P}$  is building

# Witnesses in the C-P Game



## Objective in the C-P Game

Given a play  $\rho'$  in  $\mathcal{G}'$ , there is a corresponding play  $\rho$  in  $\mathcal{G}$



If play  $\rho$  guessed to have payoff  $p$  (1)

- check that  $\text{pay}(\rho) = p$
- check that  $\rho \in \Omega_0$

Otherwise

- if  $\text{pay}(\rho) = p \in P$ , check that  $\rho \in \Omega_0$  (2)
- else check  $\text{pay}(\rho) < p \in P$  (3)

**Reachability SP game:** augment the arena with **set of satisfied objectives**  
→ checking (1-3) = **Büchi objective**

**Parity SP game:** checking (1-3) = **Boolean combination of Büchi objectives**

# C-P Game for our Running Example

