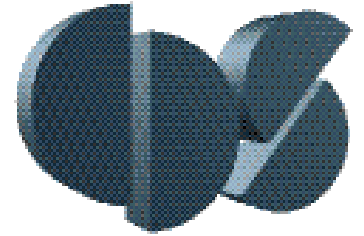




# Lecture 7

## Synthesis of Reactive Control Protocols



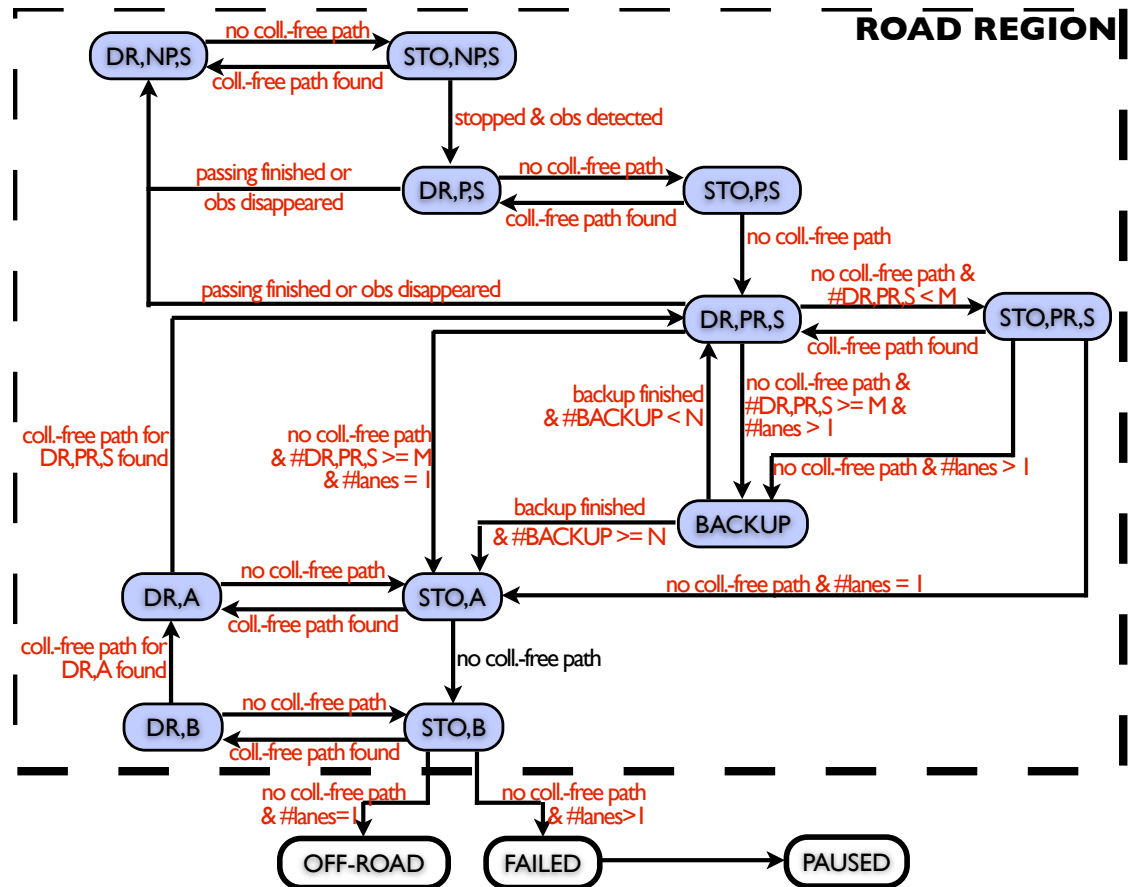
**Richard M. Murray**  
**Nok Wongpiromsarn    Ufuk Topcu**  
**California Institute of Technology**

AFRL, 25 April 2012

### Outline

- Review: networked control systems and cooperative control systems
- Asynchronous execution / group messaging systems (virtual synchrony)
- Verification of async control protocols for multi-agent, cooperative control
- Applications of model checking to Alice's actuation interface

# Alice's Logic Planner

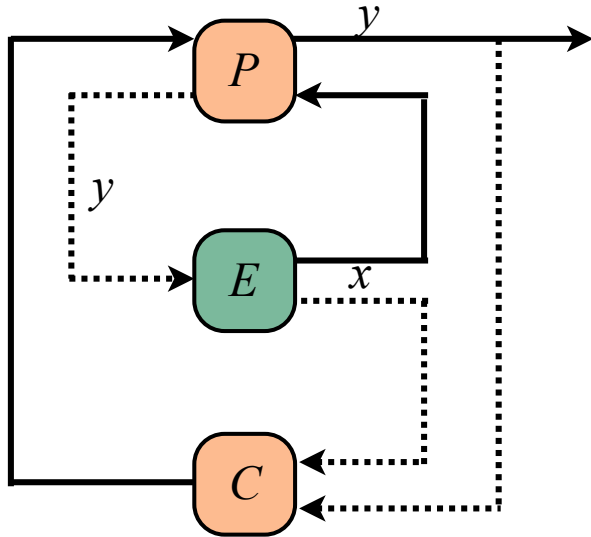


Given a specification  $\Phi$ , whether the planner is correct with respect to  $\Phi$  depends on the environment's actions (e.g., how obstacles move)

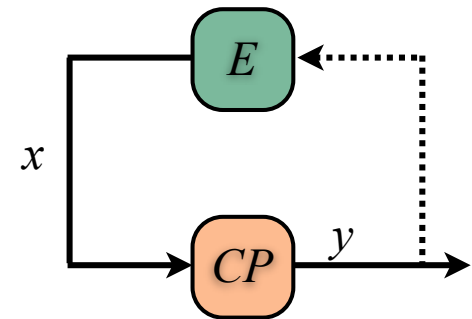
- a “correct” planner needs to ensure that  $\Phi$  is satisfied for all the possible valid behaviors of the environment

**How to design such a correct planner?**

# Open System Synthesis



An *open system* is a system whose behaviors can be affected by external influence



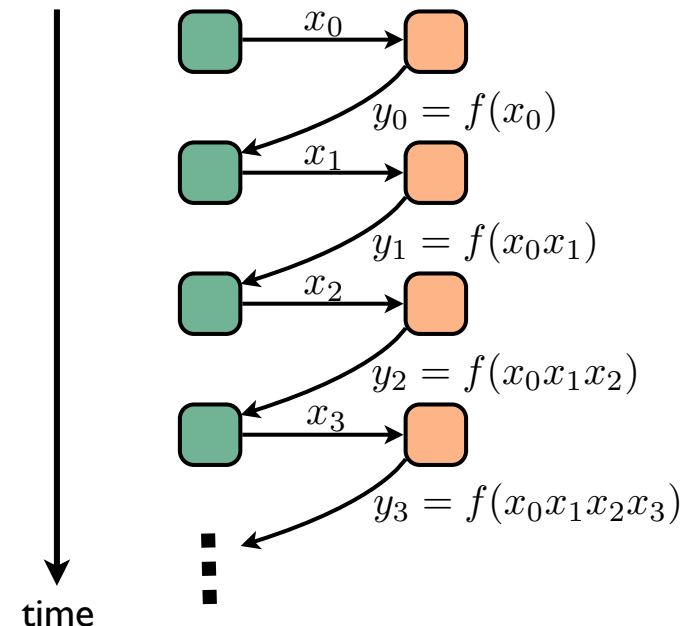
## Open (synchronous) synthesis:

Given

- a system that describes all the possible actions
  - plant actions  $y$  are controllable
  - environment actions  $x$  are uncontrollable
- a specification  $\Phi(x, y)$

find a strategy  $f(x)$  for the controllable actions which will maintain the specification against all possible adversary moves, i.e.,

$$\forall x \cdot \Phi(x, f(x))$$

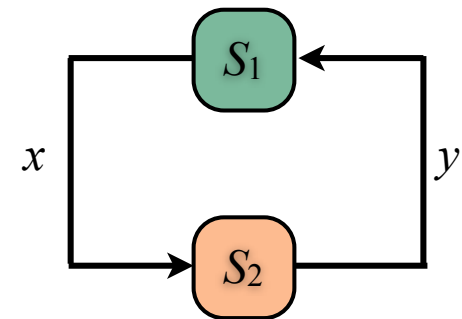


# Reactive System Synthesis

*Reactive systems* are open systems that maintain an ongoing interaction with their environment rather than producing an output on termination.

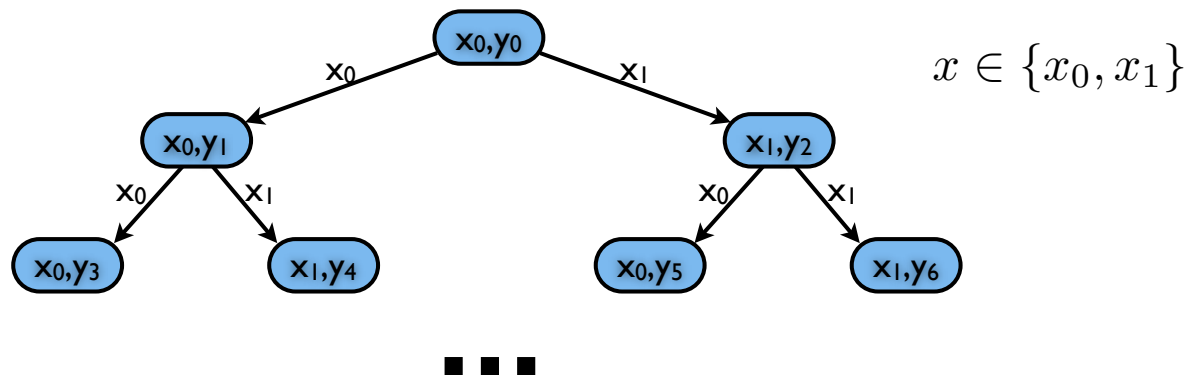
Consider the synthesis of a reactive system with input  $x$  and output  $y$ , specified by the linear temporal formula  $\Phi(x, y)$ .

- The system contains 2 components  $S_1$  (i.e., “environment”) and  $S_2$  (i.e., “reactive module”)
  - Only  $S_1$  can modify  $x$
  - Only  $S_2$  can modify  $y$
- Want to show that  $S_2$  has a *winning* strategy for  $y$  against all possible  $x$  scenarios the environment may present to it.
  - Two-person game: treat environment as adversary
    - $S_2$  does its best, by manipulating  $y$ , to maintain  $\Phi(x, y)$
    - $S_1$  does its best, by manipulating  $x$ , to falsify  $\Phi(x, y)$
- If a winning strategy for  $S_2$  exists, we say that  $\Phi(x, y)$  is *realizable*



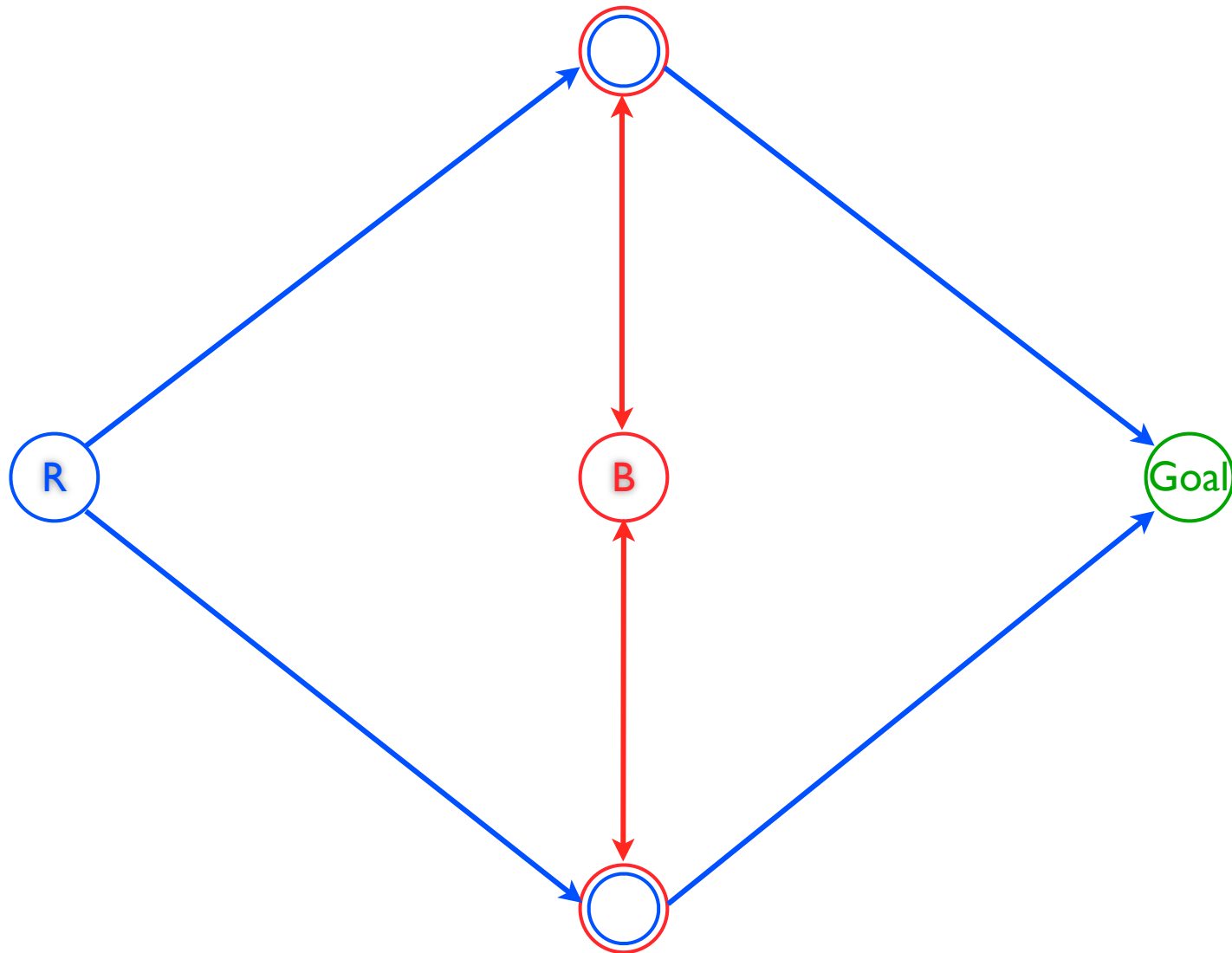
# Satisfiability $\neq$ Realizability

- Realizability should guarantee the specification against all possible (including adversarial) environment (Rosner 98)
  - To solve the problem one must find a satisfying tree where the branching represents all possible inputs



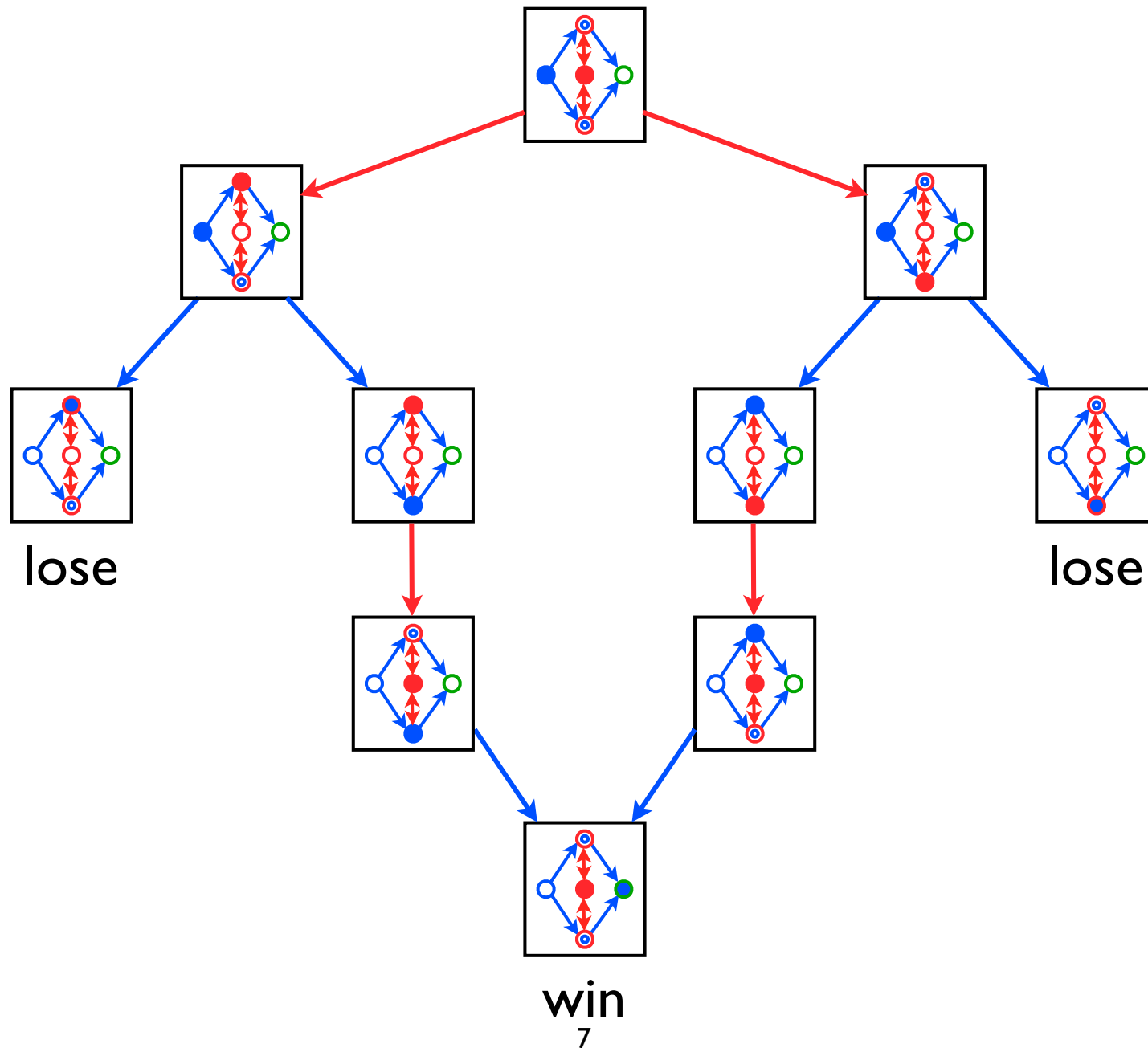
- *Satisfiability* of  $\Phi(x, y)$  only ensures that there exists at least one behavior, listing the running values of  $x$  and  $y$  that satisfies  $\Phi(x, y)$ 
  - There is a way for the plant and the environment to cooperate to achieve  $\Phi(x, y)$
- Existence of a winning strategy for  $S_2$  can be expressed by the AE-formula  $\forall x \exists y \cdot \Phi(x, y)$

# The Runner Blocker System



Runner R tries to reach Goal. Blocker B tries to intercept and stop R.

# The Runner Blocker System



# Solving Reactive System Synthesis

- Solution is typically given as the winning set
  - The winning set is the set of states starting from which there exists a strategy for  $S_2$  to satisfy the specification for all the possible behaviors of  $S_1$
  - A winning strategy can then be constructed by saving intermediate values in the winning set computation
- Worst case complexity is **double exponential**
  - Construct a nondeterministic Buchi automaton from  $\Phi(x, y) \Rightarrow$  first exponent
  - Determinize Buchi automaton into a deterministic Rabin automaton  $\Rightarrow$  second exponent
  - Follow a similar procedure as in closed system synthesis and construct the product of the system and the deterministic Rabin automaton
  - Find the set of states starting from which all the possible runs in the product automaton are accepting  $\Rightarrow$  This set can be obtained by computing the *recurrent* and the *attractor* sets
- **Special Cases of Lower Complexity**
  - For a specification of the form  $\Box p, \Diamond p, \Box \Diamond p$  or  $\Diamond \Box p$ , the controller can be synthesized in  $O(N^2)$  time where  $N$  is the size of the state space
  - Avoid translation of the formula to an automaton and determinization of the automaton

# Special Case: Satisfiability

- Transition system  $TS = (S, \text{Act}, \rightarrow, I, \text{AP}, L)$
- Specification  $\Phi = \Diamond p$
- Define the set  $WIN \triangleq \{s \in S : s \models p\}$
- Define the predecessor operator  $Pre_{\exists} : 2^S \rightarrow 2^S$  by
$$Pre_{\exists}(R) = \{s \in S : \exists r \in R \text{ s.t. } s \rightarrow r\}$$
- The set of all the states starting from which  $WIN$  is satisfiable (if the plant and the environment to cooperate) can be computed efficiently by the iteration sequence

$$R_0 = WIN$$

$$R_i = R_{i-1} \cup Pre_{\exists}(R_{i-1}), \forall i > 0$$

From **Tarski-Knaster Theorem**:

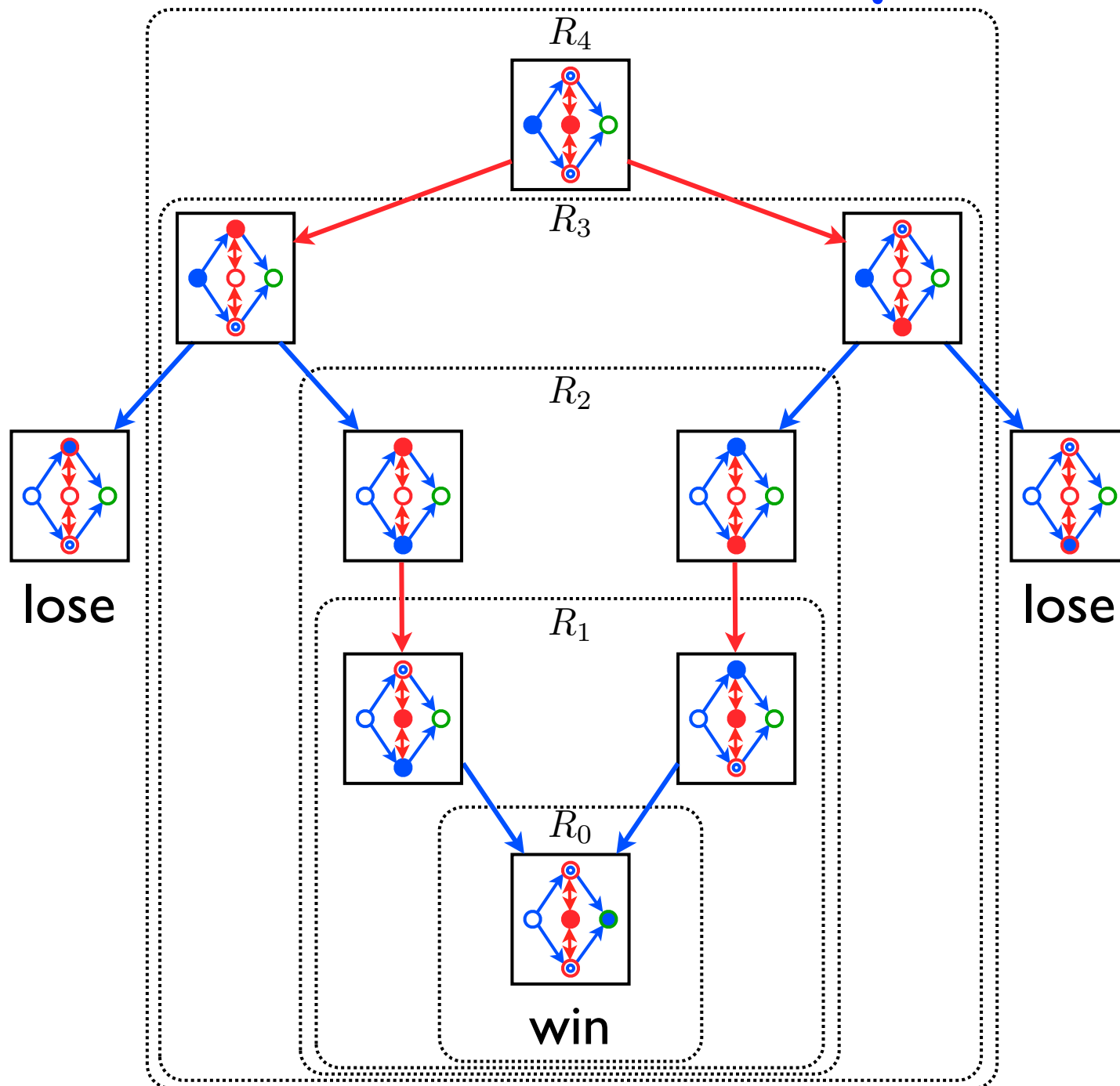
- There exists a natural number  $n$  such that  $R_n = R_{n-1}$
- Such an  $R_n$  is the minimal solution of the fix-point equation

$$R = WIN \cup Pre_{\exists}(R)$$

- The minimal solution of the above fix-point equation is denoted by

$$\mu R. (WIN \cup Pre_{\exists}(R))$$

# The Runner Blocker System



# Reachability in Adversarial Setting

- Transition system  $TS = (S, \text{Act}, \rightarrow, I, \text{AP}, L)$
- Specification  $\Phi = \Diamond p$
- Define the set  $WIN \triangleq \{s \in S : s \models p\}$
- Define the operator  $Pre_{\forall} : 2^S \rightarrow 2^S$  and  $Pre_{\exists\forall} : 2^S \rightarrow 2^S$  by

$$\begin{aligned} Pre_{\forall}(R) &= \{s \in S : \forall r \in S \text{ if } s \rightarrow r, \text{ then } r \in R\} \\ &= \text{the set of states whose all successors are in } R \end{aligned}$$

$$\begin{aligned} Pre_{\forall\exists}(R) &= Pre_{\forall}(Pre_{\exists}(R)) \\ &= \text{the set of states whose all successors} \\ &\quad \text{have at least one successor in } R \end{aligned}$$

- The set of all the states starting from which the controller can force the system into  $WIN$  can be computed efficiently by the iteration sequence

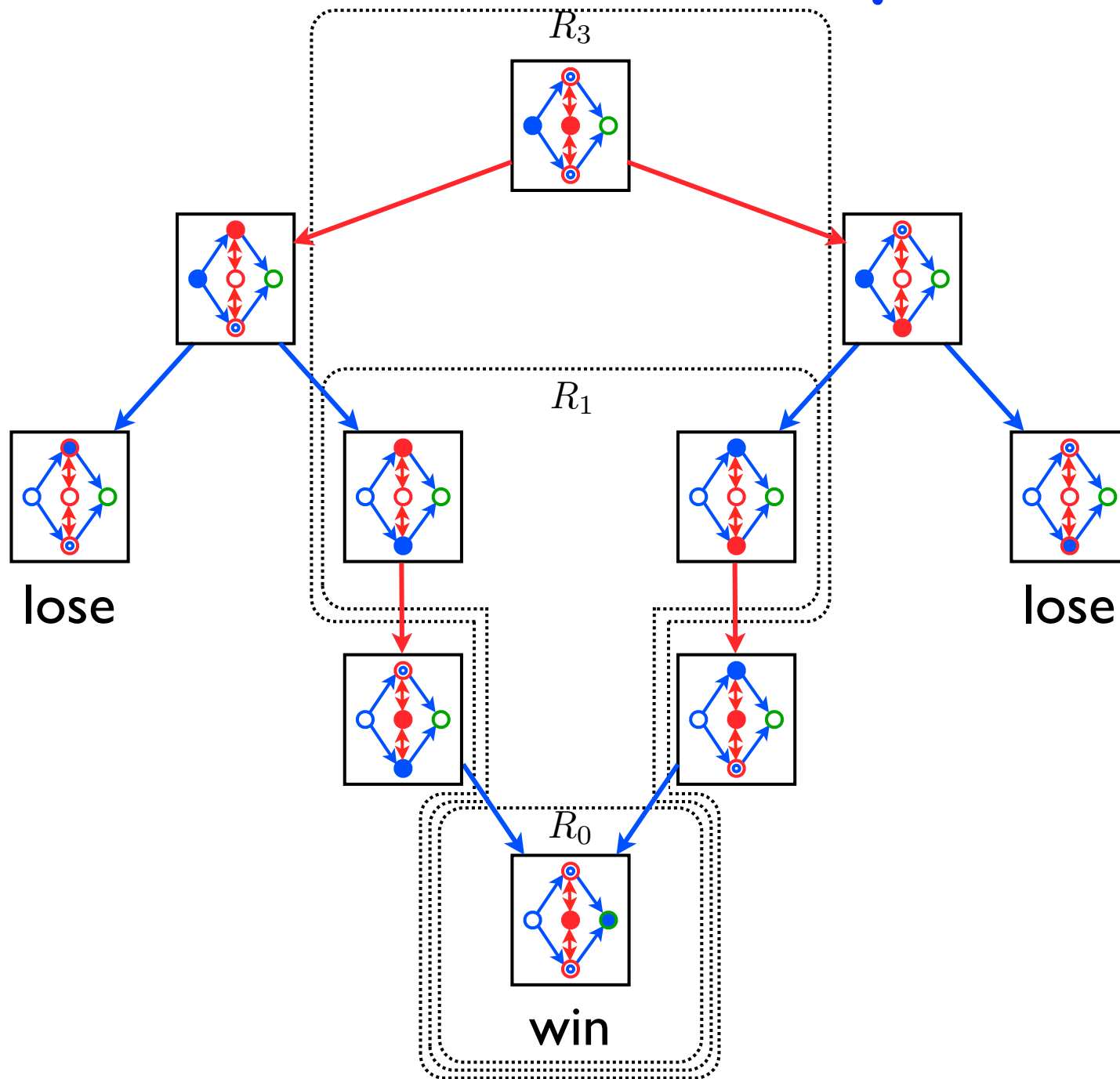
$$R_0 = WIN$$

$$R_i = R_{i-1} \cup Pre_{\forall\exists}(R_{i-1}), \forall i > 0$$

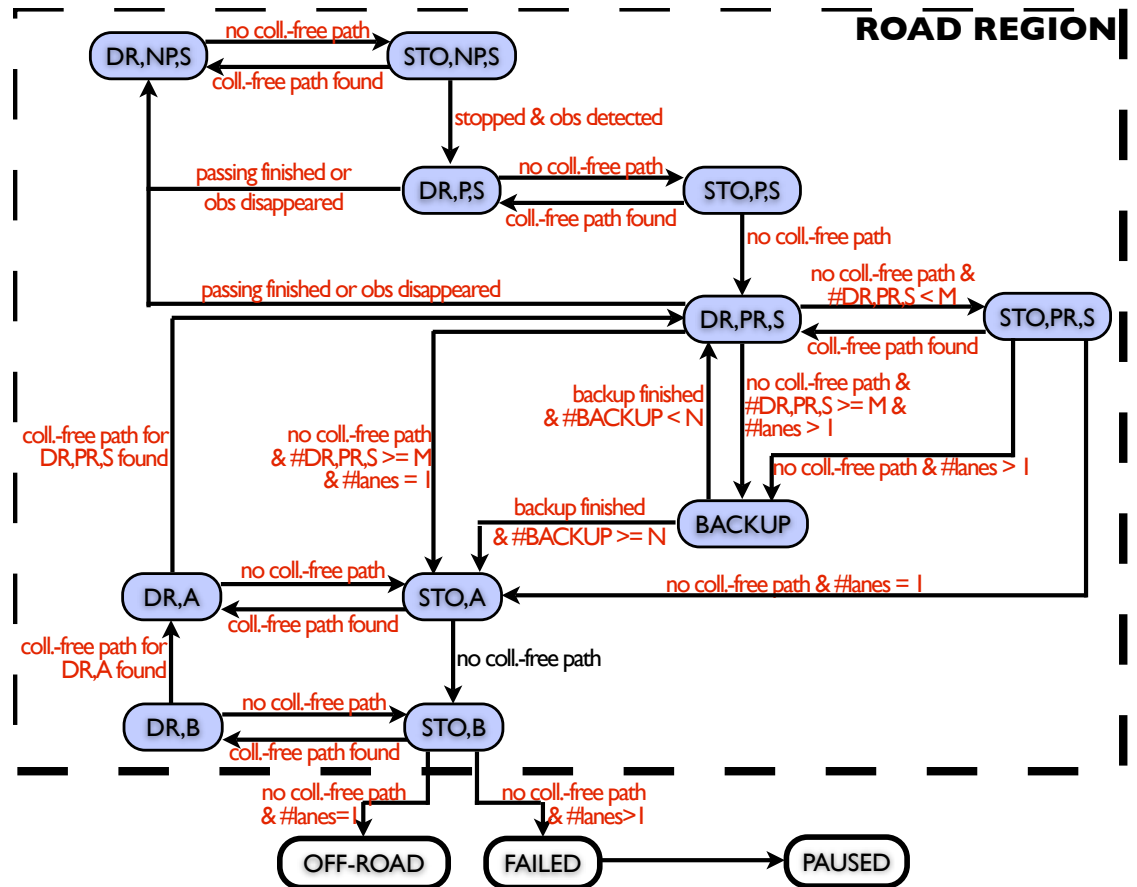
- There exists a natural number  $n$  such that  $R_n = R_{n-1}$
- Such  $R_n$  is the minimal solution of the fix-point equation  $R = WIN \cup Pre_{\forall\exists}(R)$
- The minimal solution of the above fix-point equation is denoted by

$$\mu R. (WIN \cup Pre_{\forall\exists}(R))$$

# The Runner Blocker System



# More Complicated Case



## Game Automata Approach

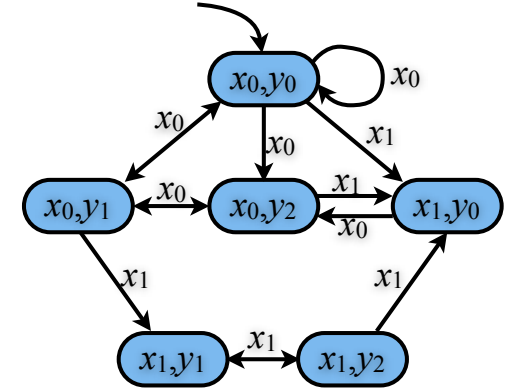
- Consider the specification as the winning condition in an infinite two-person game between input player ( $S_1$ ) and output player ( $S_2$ ).
- Decide whether player  $S_2$  has a *winning* strategy, and if this is the case construct a finite state winning strategy.

# Game Structures

A game structure is a tuple  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

- $\mathcal{V} = \{v_1, \dots, v_n\}$  is a finite set of state variables.  $\Sigma_{\mathcal{V}}$  is the set of all the possible assignments to variables in  $\mathcal{V}$
- $\mathcal{X} \subseteq \mathcal{V}$  is a set of input variables
- $\mathcal{Y} = \mathcal{V} \setminus \mathcal{X}$  is a set of output variables
- $\theta_e(\mathcal{X})$  is a proposition characterizing the initial states of the environment
- $\theta_s(\mathcal{V})$  is a proposition characterizing the initial states of the system
- $\rho_e(\mathcal{V}, \mathcal{X}')$  is a proposition characterizing the transition relation of the environment
 

primed copy of  $\mathcal{X}$  represents  
the set of next input variables
- $\rho_s(\mathcal{V}, \mathcal{X}', \mathcal{Y}')$  is a proposition characterizing the transition relation of the system
- $AP$  is a set of atomic propositions
- $L : \Sigma_{\mathcal{V}} \rightarrow 2^{AP}$  is a labeling function
- $\varphi$  is an LTL formula characterizing the winning condition



$\mathcal{V} = \{x, y\},$   
 $\mathcal{X} = \{x\}, \Sigma_{\mathcal{X}} = \{x_0, x_1\},$   
 $\mathcal{Y} = \{y\}, \Sigma_{\mathcal{Y}} = \{y_0, y_1, y_2\},$   
 $x_0 \models \theta_e, x_1 \not\models \theta_e,$   
 $(x_0, y_0) \models \theta_s,$   
 $(x_i, y_j) \not\models \theta_s, \forall i, j \neq 0,$   
 $((x_0, y_i), x_j) \models \rho_e, \forall i, j,$   
 $((x_1, y_0), x_0) \models \rho_e,$   
 $((x_1, y_0), x_1) \not\models \rho_e,$   
 $((x_1, y_i), x_0) \not\models \rho_e, \forall i \in \{1, 2\},$   
 $((x_1, y_i), x_1) \models \rho_e, \forall i \in \{1, 2\},$   
 $((x_0, y_0), x_0, y_i) \models \rho_s, \forall i,$   
 $((x_0, y_0), x_1, y_0) \models \rho_s,$   
 $((x_0, y_0), x_1, y_i) \not\models \rho_s, \forall i \neq 0,$   
 $\dots$

# Autonomous Car Example



**Game Structure**  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

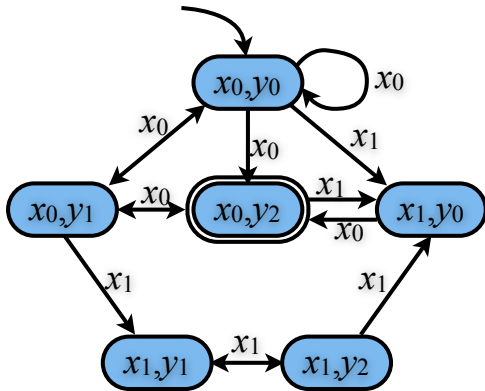
- $\mathcal{X}$  (environment): obstacles, other cars, pedestrians
- $\mathcal{Y}$  (plant): vehicle state (drive VS stop, passing?, reversing?, etc)
- $\theta_e$  describes the valid initial states of the environment, e.g., where obstacles can be
- $\theta_s$  describes the valid initial states of the vehicle, e.g., the stop state
- $\rho_e$  describes how obstacles may move
- $\rho_s$  describes the valid transitions of the vehicle state
- $\varphi$  describes the winning condition, e.g., vehicle does not get stuck

# Plays

Game structure  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

- A *play* of  $G$  is a maximal sequence of states  $\sigma = s_0 s_1 \dots$  satisfying  $s_0 \models \theta_e \wedge \theta_s$  and  $(s_j, s_{j+1}) \models \rho_e \wedge \rho_s, \forall j \geq 0$ .  
infinite or the last state in the sequence has no valid successor
  - Initially, the environment chooses an assignment  $s_{\mathcal{X}} \in \Sigma_{\mathcal{X}}$  such that  $s_{\mathcal{X}} \models \theta_e$  and the system chooses an assignment  $s_{\mathcal{Y}} \in \Sigma_{\mathcal{Y}}$  such that  $(s_{\mathcal{X}}, s_{\mathcal{Y}}) \models \theta_e \wedge \theta_s$ .
  - From a state  $s_j$ , the environment chooses an input  $s_{\mathcal{X}} \in \Sigma_{\mathcal{X}}$  such that  $(s_j, s_{\mathcal{X}}) \models \rho_e$  and the system chooses an output  $s_{\mathcal{Y}} \in \Sigma_{\mathcal{Y}}$  such that  $(s, s_{\mathcal{X}}, s_{\mathcal{Y}}) \models \rho_s$ .
- A play  $\sigma$  is *winning for the system* if either
  - $\sigma = s_0 s_1 \dots s_n$  is finite and  $(s_n, s_{\mathcal{X}}) \not\models \rho_e, \forall s_{\mathcal{X}} \in \Sigma_{\mathcal{X}}$ , or
  - $\sigma$  is infinite and  $\sigma \models \varphi$ .

Otherwise  $\sigma$  is *winning for the environment*.



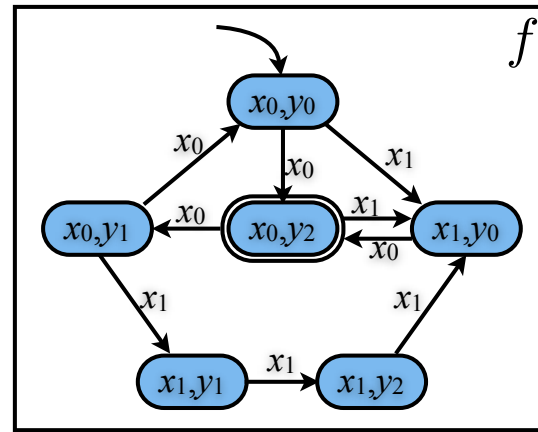
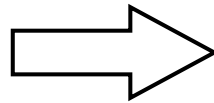
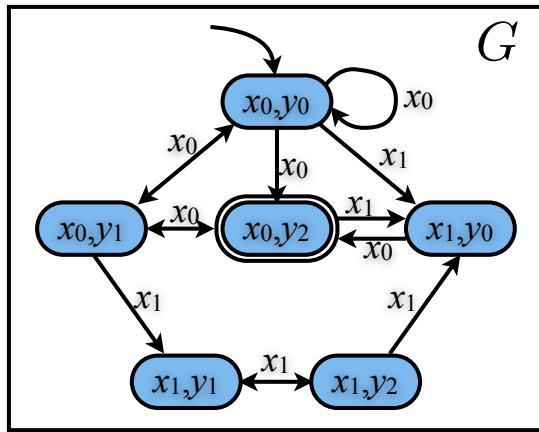
$$\varphi = \Box \Diamond (x = x_0 \wedge y = y_2)$$

- $\sigma = ((x_0, y_0), (x_0, y_2), (x_0, y_1))^\omega$  is winning for the system
- $\sigma = ((x_0, y_0))^\omega$  is winning for the environment

# Strategies

Game structure  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

- A *strategy for the system* is a function  $f : M \times \Sigma_{\mathcal{V}} \times \Sigma_{\mathcal{X}} \rightarrow M \times \Sigma_{\mathcal{Y}}$  such that for all  $s \in \Sigma_{\mathcal{V}}, s_{\mathcal{X}} \in \Sigma_{\mathcal{X}}, m \in M$ , if  $f(m, s, s_{\mathcal{X}}) = (m', s_{\mathcal{Y}})$  and  $(s, s_{\mathcal{X}}) \models \rho_e$ , then  $(s, s_{\mathcal{X}}, s_{\mathcal{Y}}) \models \rho_s$ .
- A play  $\sigma = s_0 s_1 \dots$  is *compliant* with strategy  $f$  if  $f(m_i, s_i, s_{i+1}|_{\mathcal{X}}) = (m_{i+1}, s_{i+1}|_{\mathcal{Y}}), \forall i$ .
- A strategy  $f$  is *winning for the system* from state  $s \in \Sigma_{\mathcal{V}}$  if all plays that start from  $s$  and are compliant with  $f$  are winning for the system. If such a winning strategy exists, we call  $s$  a *winning state for the system*.



Is  $f$  winning for the system?

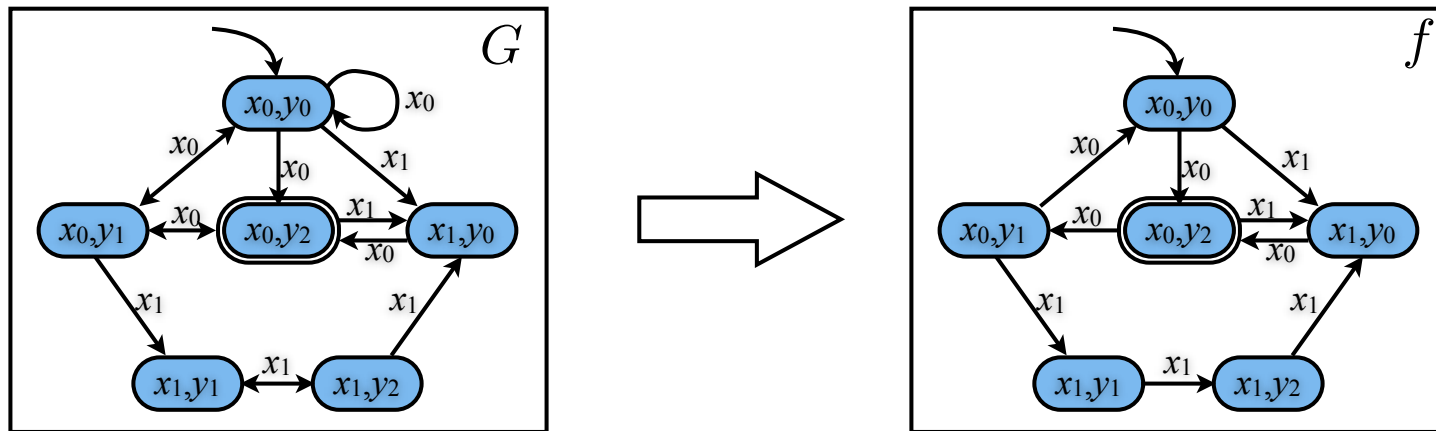
$$\begin{aligned} f(m, (x_0, y_0), x_0) &= (m, y_2) \\ f(m, (x_0, y_0), x_1) &= (m, y_0) \\ f(m, (x_0, y_1), x_0) &= (m, y_0) \\ f(m, (x_0, y_1), x_1) &= (m, y_1) \end{aligned}$$

$$\begin{aligned} f(m, (x_0, y_2), x_0) &= (m, y_1) \\ f(m, (x_0, y_2), x_1) &= (m, y_0) \\ f(m, (x_1, y_0), x_0) &= (m, y_2) \\ f(m, (x_1, y_0), x_1) &= (m, y_2) \end{aligned}$$

$$\begin{aligned} f(m, (x_1, y_1), x_0) &= (m, y_2) \\ f(m, (x_1, y_1), x_1) &= (m, y_2) \\ f(m, (x_1, y_2), x_0) &= (m, y_2) \\ f(m, (x_1, y_2), x_1) &= (m, y_0) \end{aligned}$$

# Winning Games

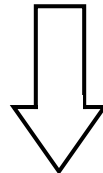
A game structure  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$  is *winning for the system* if for each  $s_{\mathcal{X}} \in \Sigma_{\mathcal{X}}$  such that  $s_{\mathcal{X}} \models \theta_e$ , there exists  $s_{\mathcal{Y}} \in \Sigma_{\mathcal{Y}}$  such that  $(s_{\mathcal{X}}, s_{\mathcal{Y}}) \models \theta_s$  and  $(s_{\mathcal{X}}, s_{\mathcal{Y}})$  is a winning state for the system



$(x_0, y_0)$  is a winning state for the system

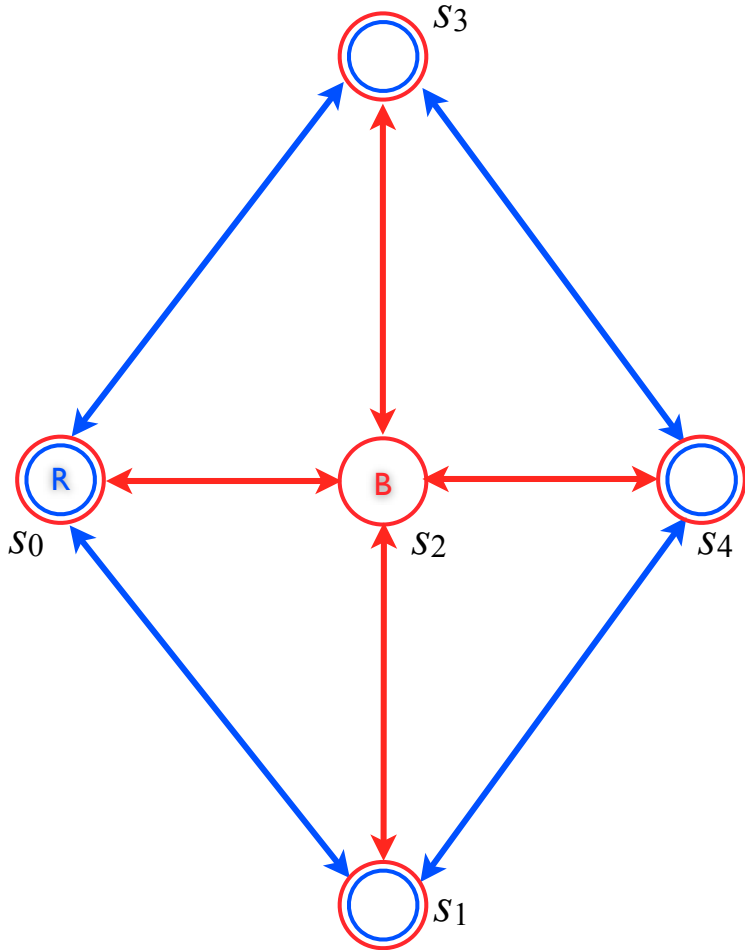
$x_0 \models \theta_e$  but  $x_1 \not\models \theta_e$

$(x_0, y_0) \models \theta_s$



$G$  is winning for the system

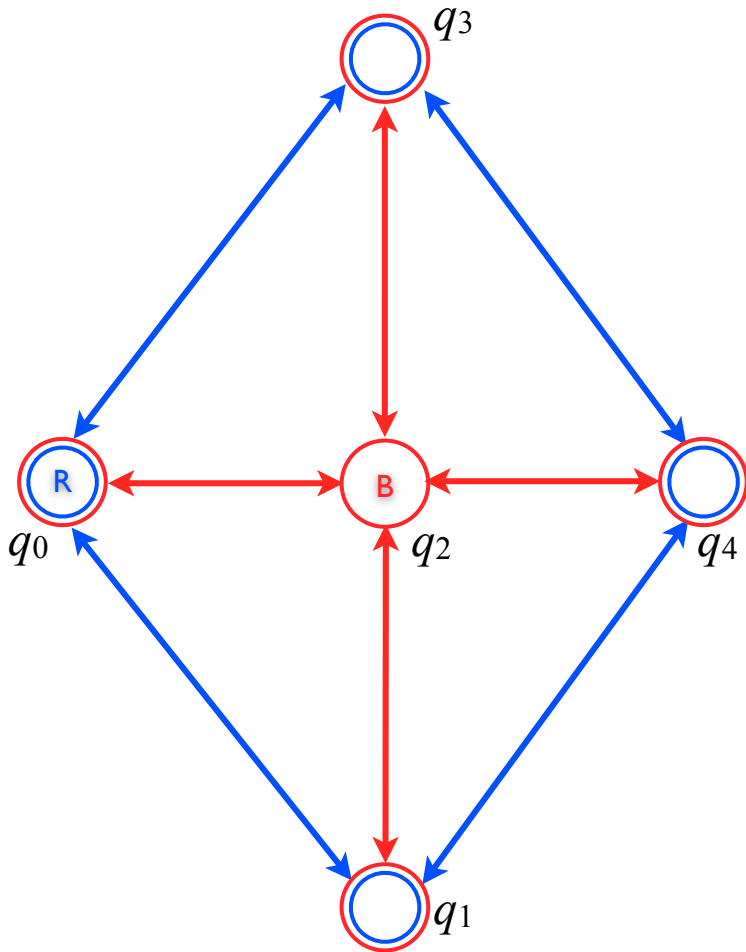
# Runner Blocker Example



**Game Structure**  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

- $\mathcal{X} := \{x\}, \Sigma_{\mathcal{X}} = \{s_0, s_1, s_2, s_3, s_4\}$
- $\mathcal{Y} := \{y\}, \Sigma_{\mathcal{Y}} = \{s_0, s_1, s_3, s_4\}$
- $\theta_e := (x = s_2)$
- $\theta_s := (y = s_0)$
- $\rho_e := ((x = s_2) \implies (x' \neq s_2)) \wedge ((x \neq s_2) \implies (x' = s_2))$
- $\rho_s := ((y = s_0 \vee y = s_4) \implies (y' = s_1 \vee y' = s_3)) \wedge ((y = s_1 \vee y = s_3) \implies (y' = s_0 \vee y' = s_4)) \wedge (y' \neq x')$
- $\varphi$  describes the winning condition, e.g.,  $\diamond(y = s_4)$

# Runner Blocker Example



**Play:** An infinite sequence  $\sigma = s_0 s_1 \dots$  of system (blocker + runner) states such that  $s_0$  is a valid initial state and  $(s_j, s_{j+1})$  satisfies the transition relation of the blocker and the runner

**Strategy:** A function that gives the next runner state, given a finite number of previous system states of the current play, the current system state and the next blocker state

**Winning state:** A state starting from which there exists a strategy for the runner to satisfy the winning condition for all the possible behaviors of the blocker

**Winning game:** For any valid initial blocker state  $s_x$ , there exists a valid initial runner state  $s_y$  such that  $(s_x, s_y)$  is a winning state

**Solving game:** Identify the set of winning states

# Solving Game Structures

## General solutions are hard

- Worst case complexity is double exponential (roughly in number of states)

## Special cases are easier

- For a specification of the form  $\Box p$ ,  $\Diamond p$ ,  $\Box\Diamond p$  or  $\Diamond\Box p$ , the controller can be synthesized in  $O(N^2)$  time where  $N$  is the size of the state space

## Another special case: GR(1) formulas

$$\varphi = \underbrace{(\Box\Diamond p_1 \wedge \dots \wedge \Box\Diamond p_m)}_{\varphi_e} \implies \underbrace{(\Box\Diamond q_1 \wedge \dots \wedge \Box\Diamond q_n)}_{\varphi_s}$$

**Thm** (Piterman, Sa'ar, Pnueli, 2007) A game structure  $G$  with a GR(1) winning condition can be solved by a symbolic algorithm in time proportional to  $nm|\Sigma_V|^3$

## More useful form:

$$\varphi = \underbrace{(\psi_{init}^e)}_{\text{assumptions on initial condition}} \wedge \underbrace{(\Box\psi_s^e \wedge \bigwedge_{i \in I_f} \Box\Diamond\psi_{f,i}^e)}_{\text{assumptions on environment}} \implies \underbrace{(\psi_{init}^s \wedge \Box\psi_s^s \wedge \bigwedge_{i \in I_g} \Box\Diamond\psi_{g,i}^s)}_{\text{desired behavior}}$$

- Can show (tomorrow) that this can be “converted” to GR(1) form

# Solving Reachability Games

- Game structure  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

- For a proposition  $p$ , let

$$[[p]] = \{s \in \Sigma_{\mathcal{V}} \mid s \models p\}$$

- For a set  $R$ , let

$$[[\odot R]] = \left\{ s \in \Sigma_{\mathcal{V}} \mid \forall s'_{\mathcal{X}} \in \Sigma_{\mathcal{X}}, (s, s'_{\mathcal{X}}) \models \rho_e \Rightarrow \exists s'_{\mathcal{Y}} \in \Sigma_{\mathcal{Y}} \text{ s.t. } (s, s'_{\mathcal{X}}, s'_{\mathcal{Y}}) \models \rho_s \text{ and } (s'_{\mathcal{X}}, s'_{\mathcal{Y}}) \in R \right\}$$

similar to the  $Pre_{\forall\exists}$  operator we saw earlier

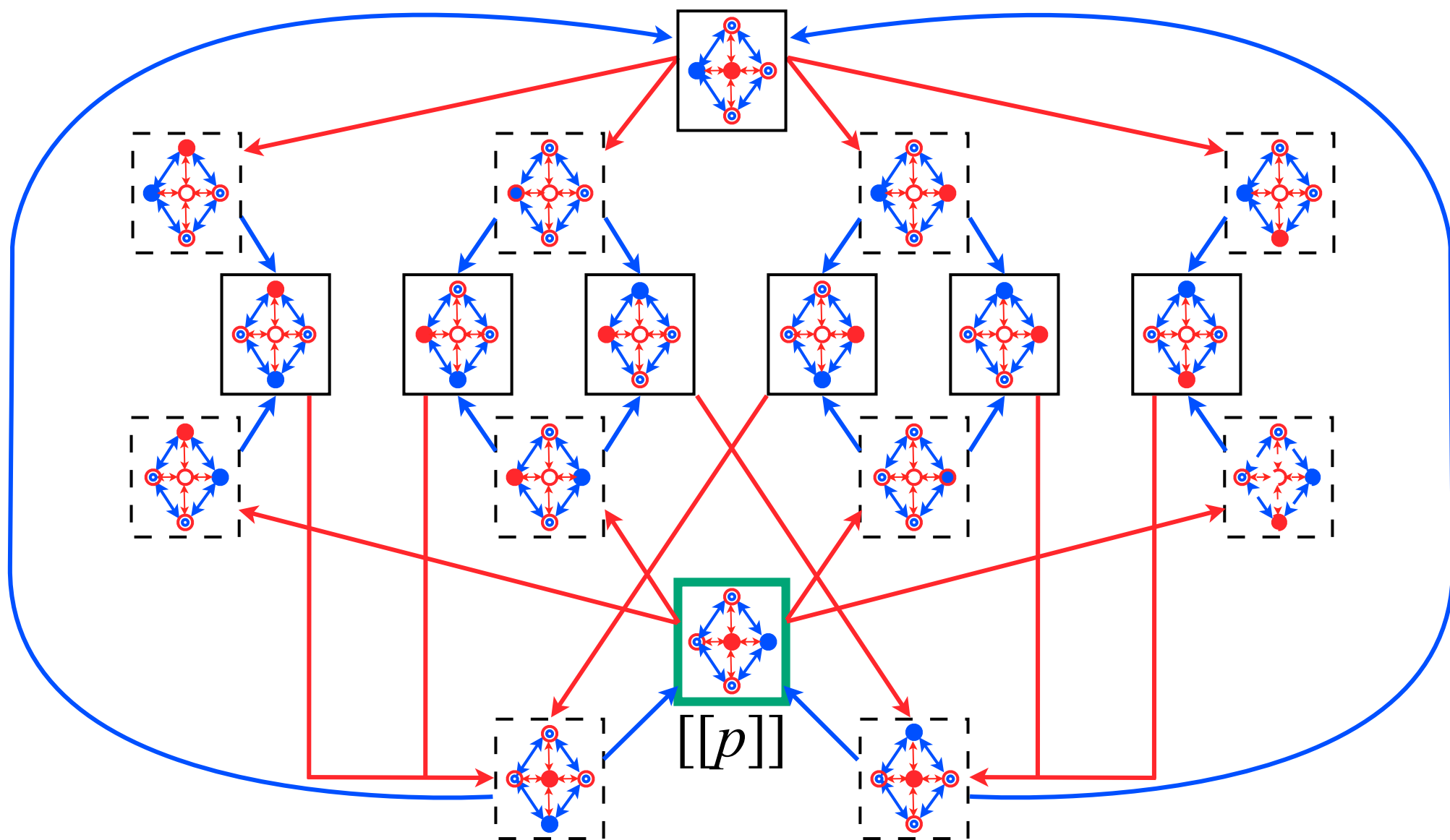
- Reachability game:  $\varphi = \diamond p$
- The set of winning states can be computed efficiently by the iteration sequence

$$\begin{aligned} R_0 &= \emptyset \\ R_{i+1} &= [[p]] \cup [[\odot R_i]], \forall i \geq 0 \end{aligned}$$

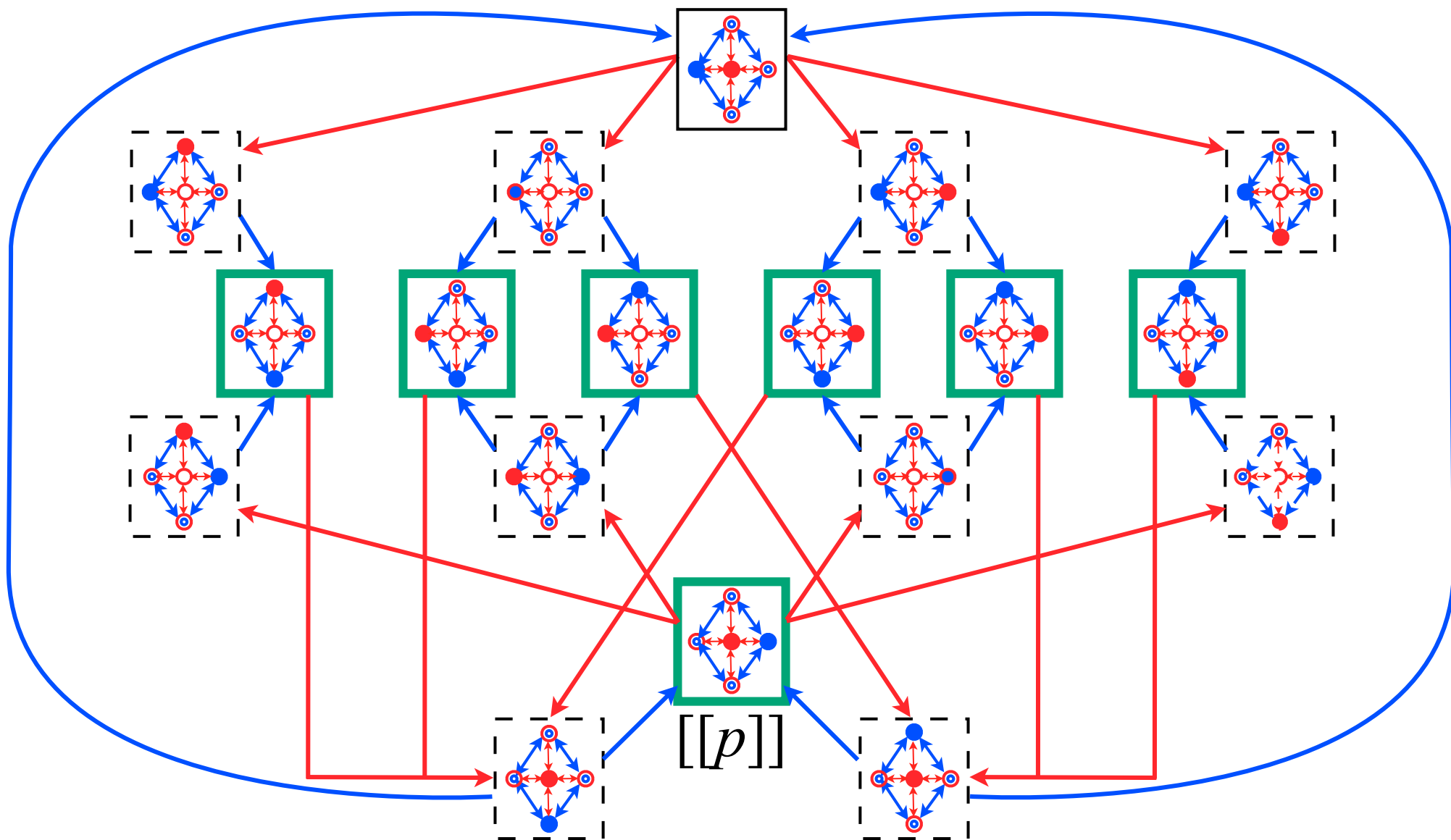
- $R_{i+1}$  is the set of states starting from which the system can force the play to reach a state satisfying  $p$  within  $i$  steps
- There exists a natural number  $n$  such that  $R_n = R_{n-1}$
- Such  $R_n$  is the minimal solution of the fix-point equation  $R = [[p]] \cup [[\odot R]]$
- In  $\mu$ -calculus, the minimal solution of the above fix-point equation is denoted by  $\mu R(p \vee \odot R)$

least fixpoint

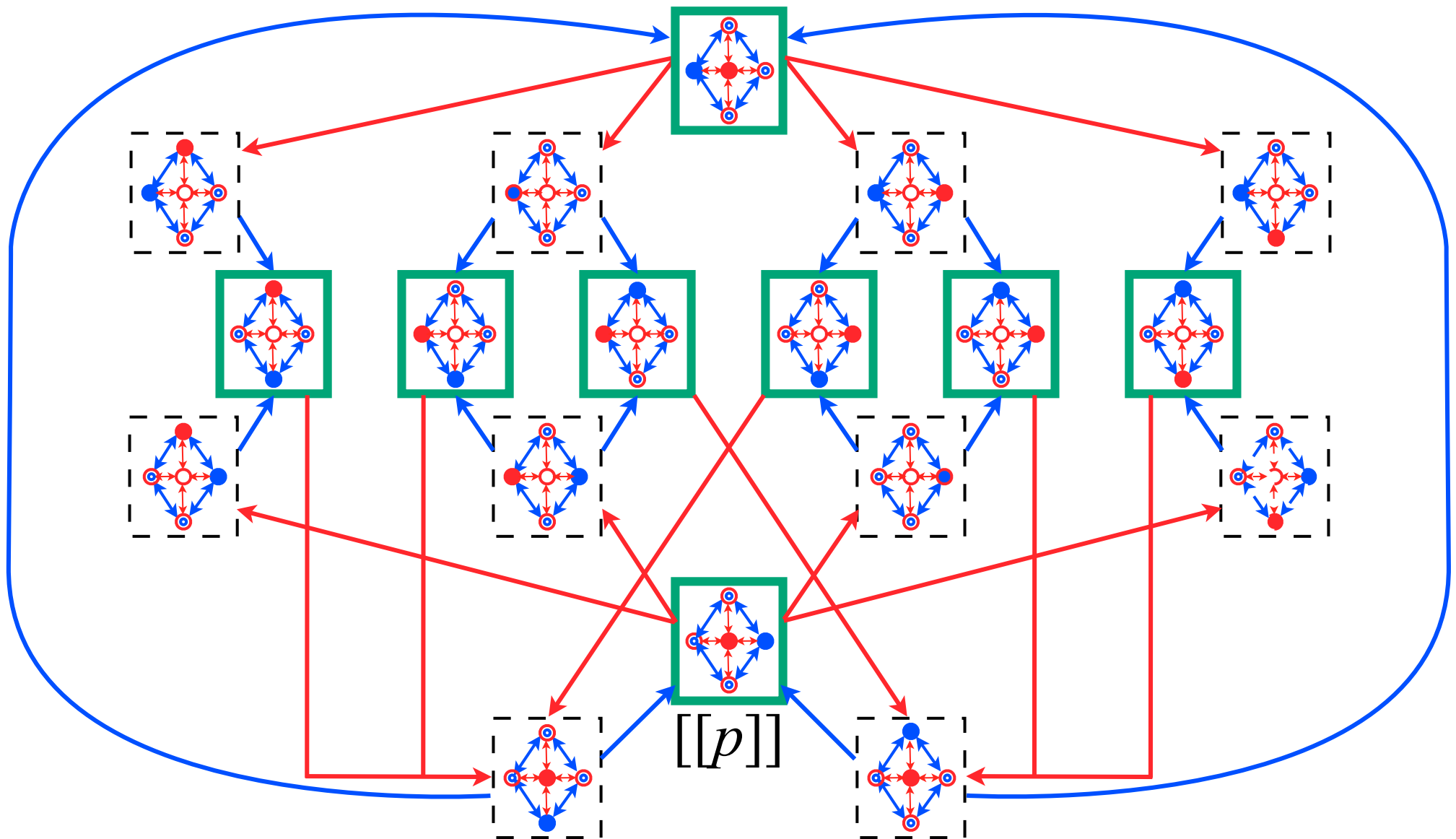
# Runner Blocker Example: $R_1$



# Runner Blocker Example: $R_2$



# Runner Blocker Example: $R_3 = R_4 = \dots$



# Solving Safety Games

- Game structure  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

- For a proposition  $p$ , let

$$[[p]] = \{s \in \Sigma_{\mathcal{V}} \mid s \models p\}$$

- For a set  $R$ , let

$$[[\odot R]] = \left\{ s \in \Sigma_{\mathcal{V}} \mid \forall s'_{\mathcal{X}} \in \Sigma_{\mathcal{X}}, (s, s'_{\mathcal{X}}) \models \rho_e \Rightarrow \exists s'_{\mathcal{Y}} \in \Sigma_{\mathcal{Y}} \text{ s.t. } (s, s'_{\mathcal{X}}, s'_{\mathcal{Y}}) \models \rho_s \text{ and } (s'_{\mathcal{X}}, s'_{\mathcal{Y}}) \in R \right\}$$

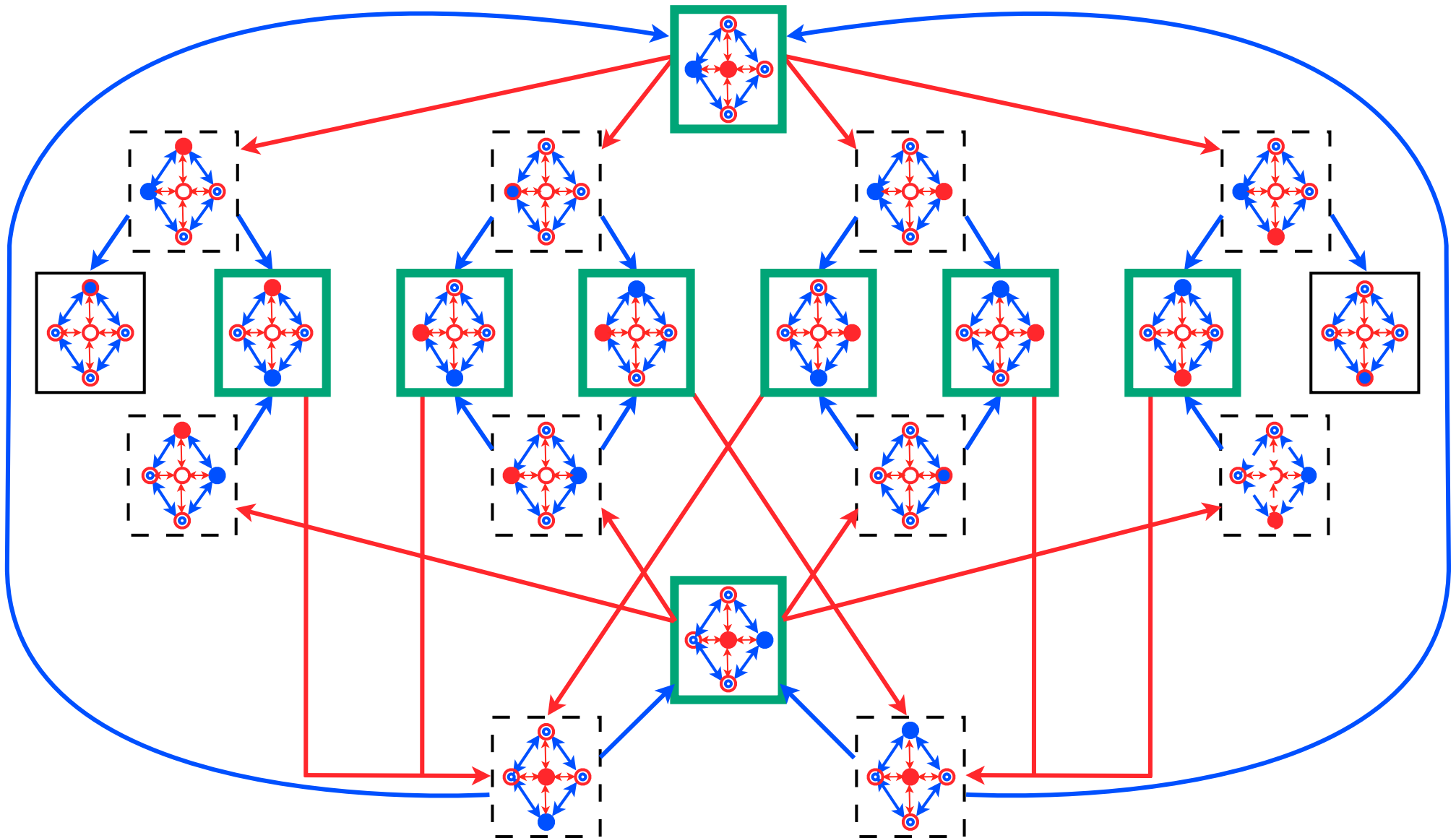
- Safety game:  $\varphi = \Box p$
- The set of winning states can be computed efficiently by the iteration sequence

$$\begin{aligned} R_0 &= \Sigma_{\mathcal{V}} \\ R_{i+1} &= [[p]] \cap [[\odot R_i]], \forall i \geq 0 \end{aligned}$$

- $R_{i+1}$  is the set of states starting from which the system can force the play to stay in states satisfying  $p$  for  $i$  steps
- There exists a natural number  $n$  such that  $R_n = R_{n-1}$
- Such  $R_n$  is the maximal solution of the fix-point equation  $R = [[p]] \cap [[\odot R]]$
- In  $\mu$ -calculus, the minimal solution of the above fix-point equation is denoted by  $\nu R(p \wedge \odot R)$

 greatest fixpoint

# Runner Blocker Example: $R_1 = R_2 = \dots$



# Solving Games

Game structure  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

$\varphi$	The set of winning states for the system
$\Diamond p$	$\mu X(p \vee \Diamond X)$
$\Box p$	$\nu X(p \wedge \Diamond X)$
$\Box \Diamond p$	$\nu X \mu Y((p \wedge \Diamond X) \vee \Diamond Y)$

- $\nu X(p \wedge \Diamond X)$  is the largest set  $S$  of states such that
  - all the states in  $S$  satisfy  $p$ , and
  - starting from a state in  $S$ , the system can force the play to transition to a state in  $S$
- $\nu X \mu Y((p \wedge \Diamond X) \vee \Diamond Y)$  is the set of state starting from which the system can force the play to satisfy  $p$  infinitely often
  - The disjunction and  $\mu Y$  operators ensure that the system is in a state where it can force the play to reach a state satisfying  $p$
  - The conjunction and the  $\nu X$  operators ensure that the above statement is true at all time

# Games and Realizability

Game structure  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$

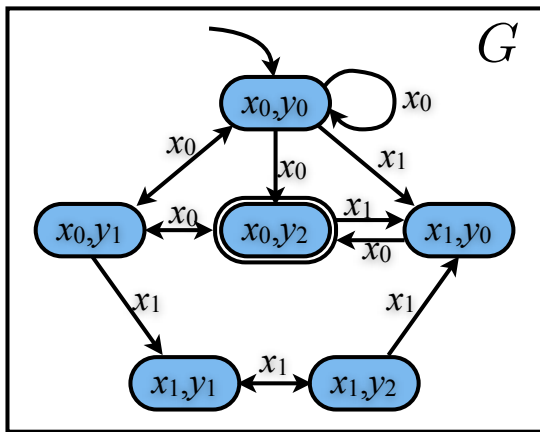
The system wins in  $G$  iff the specification

$$\psi = (\theta_e \implies \theta_s) \wedge (\theta_e \implies \Box((\Box\rho_e) \implies \rho_s)) \wedge ((\theta_e \wedge \Box\rho_e) \implies \varphi)$$

is realizable.

Given an LTL specification  $\psi$ , we construct  $G$  as follows

- $\theta_e$  and  $\theta_s$  include the non-temporal specification parts of  $\psi$
- $\rho_e$  and  $\rho_s$  include the local limitations on the next values of variables in  $\mathcal{X}$  and  $\mathcal{Y}$
- $\varphi$  includes all the remaining properties in  $\psi$  that are not included in  $\theta_e, \theta_s, \rho_e$  and  $\rho_s$



$$X_i \triangleq (x = x_i), \quad Y_i \triangleq (y = y_i), \quad X'_i \triangleq (x' = x_i), \quad Y'_i \triangleq (y' = y_i)$$

$$\theta_e \triangleq X_0, \quad \theta_s \triangleq Y_0$$

$$\rho_e \triangleq ((X_1 \wedge Y_0) \implies X'_0) \wedge ((X_1 \wedge Y_1) \implies X'_1) \wedge ((X_1 \wedge Y_2) \implies X'_1)$$

$$\begin{aligned} \rho_s \triangleq & ((X_0 \wedge Y_0 \wedge X'_0) \implies (Y'_1 \vee Y'_2)) \wedge ((X_0 \wedge Y_0 \wedge X'_1) \implies (Y'_0)) \wedge \\ & ((X_0 \wedge Y_1 \wedge X'_0) \implies (Y'_0 \vee Y'_2)) \wedge ((X_0 \wedge Y_1 \wedge X'_1) \implies (Y'_1)) \wedge \\ & ((X_0 \wedge Y_2 \wedge X'_0) \implies Y'_1) \wedge ((X_0 \wedge Y_2 \wedge X'_1) \implies Y'_0) \wedge \\ & ((X_1 \wedge Y_0 \wedge X'_0) \implies Y'_2) \wedge ((X_1 \wedge Y_1 \wedge X'_1) \implies Y'_2) \wedge \\ & ((X_1 \wedge Y_2 \wedge X'_1) \implies (Y'_0 \vee Y'_1)) \end{aligned}$$

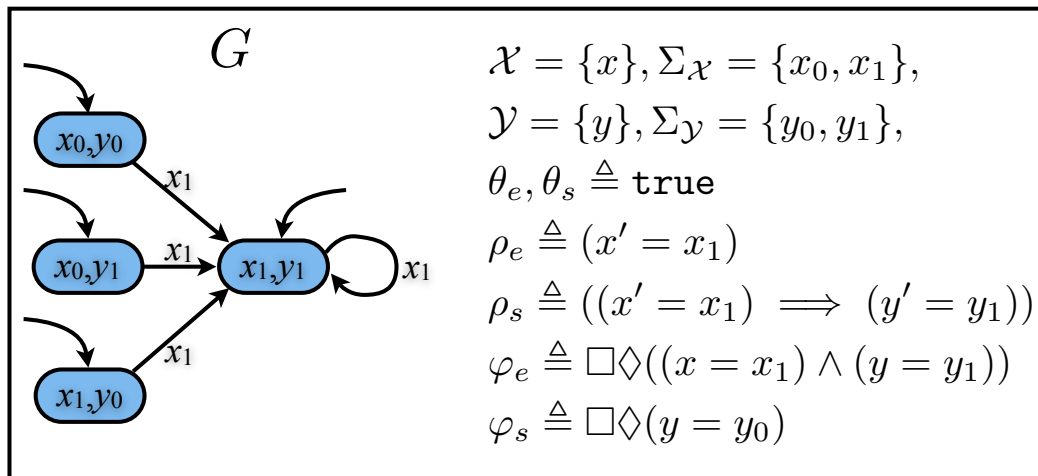
$$\varphi \triangleq \Box\Diamond(X_0 \wedge Y_2)$$

# Games and Realizability

More intuitive specification

$$\psi' = (\theta_e \wedge \Box \rho_e \wedge \varphi_e) \implies (\theta_s \wedge \Box \rho_s \wedge \varphi_s)$$

- Fulfillment of the system safety depends on the liveness of the environment
  - The system may violate its safety if it ensures that the environment cannot fulfill its liveness
- $\psi$  implies  $\psi'$ 
  - If  $\psi$  is realizable, a controller for  $\psi$  is also a controller for  $\psi'$  (but not vice versa)
  - If the system wins in  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi_e \implies \varphi_s)$ , then  $\psi'$  is realizable (but not vice versa)
- By adding extra output variables that represent the memory of whether the system or the environment violate their initial requirements or their safety requirements, we can construct a game  $G'$  such that  $G'$  is won by the system iff  $\psi'$  is realizable



- $\psi'$  is realizable
  - The system always picks  $y = y_0$
- $\psi$  is not realizable
- The system does not win in  $G$

# General Reactivity(1) Games

GR(I) game is a game  $G = (\mathcal{V}, \mathcal{X}, \mathcal{Y}, \theta_e, \theta_s, \rho_e, \rho_s, AP, L, \varphi)$  with the winning condition

$$\varphi = \underbrace{(\Box \Diamond p_1 \wedge \dots \wedge \Box \Diamond p_m)}_{\varphi_e} \implies \underbrace{(\Box \Diamond q_1 \wedge \dots \wedge \Box \Diamond q_n)}_{\varphi_s}$$

The winning states in a GR(I) game can be computed using the fixpoint expression

$$\nu \begin{bmatrix} Z_1 \\ Z_2 \\ \vdots \\ Z_n \end{bmatrix} \left[ \begin{array}{c} \mu Y \left( \bigvee_{i=1}^m \nu X \left( (q_1 \wedge \Diamond Z_2) \vee \Diamond Y \vee (\neg p_i \wedge \Diamond X) \right) \right) \\ \mu Y \left( \bigvee_{i=1}^m \nu X \left( (q_2 \wedge \Diamond Z_3) \vee \Diamond Y \vee (\neg p_i \wedge \Diamond X) \right) \right) \\ \vdots \\ \mu Y \left( \bigvee_{i=1}^m \nu X \left( (q_n \wedge \Diamond Z_1) \vee \Diamond Y \vee (\neg p_i \wedge \Diamond X) \right) \right) \end{array} \right]$$

- $\mu Y \nu X (\Diamond Y \vee (\neg p_i \wedge \Diamond X))$  characterizes the set of states from which the system can force the play to stay indefinitely in  $\neg p_i$  states
- The two outer fixpoints make sure that the system wins from the set  $q_j \wedge \Diamond Z_{j \oplus 1} \vee \Diamond Y$ 
  - The disjunction and  $\mu Y$  operators ensure that the system is in a state where it can force the play to reach a  $q_j \wedge \Diamond Z_{j \oplus 1}$  state in a finite number of steps
  - The conjunction and  $\nu Z_j$  operators ensure that after visiting  $q_j$ , we can loop and visit  $q_{j \oplus 1}$

# Lecture Schedule

	Tue	Wed	Thu
8:30	L1: Intro to Protocol-Based Control Systems	<b>Computer Lab 1</b> Spin	L8: Receding Horizon Temporal Logic Planning
10:30	L2: Automata Theory	L5: Verification of Control Protocols	<b>Computer Lab 2</b> TuLiP
12:00	Lunch	Lunch	Lunch
13:30	L3: Linear Temporal Logic	L6: Hybrid Systems Verification	L9: Extensions, Applications and Open Problems
15:30	L4: Model Checking and Logic Synthesis	L7: Synthesis of Reactive Control Protocols	

<http://www.cds.caltech.edu/~murray/wiki/afrlcourse2012>

# Extracting GR(1) Strategies

The intermediate values in the computation of the fixpoint can be used to compute a strategy, represented by a finite transition system, for a GR(1) game.

This strategy does one of the followings

- Iterates over strategies  $f_1, \dots, f_n$  where  $f_j$  ensures that the play reaches a  $q_j$  state
- Eventually uses a fixed strategy ensuring that the play does not satisfy one of the liveness assumptions  $p_j$

**Complexity:** A game structure  $G$  with a GR(1) winning condition can be solved by a symbolic algorithm in time proportional to  $nm|\Sigma_V|^3$

# Extensions

The algorithm for solving GR(I) game can be applied to any game with the winning condition of the form

$$\varphi = \underbrace{(\Box\Diamond p_1 \wedge \dots \wedge \Box\Diamond p_m)}_{\varphi_e} \implies \underbrace{(\Box\Diamond q_1 \wedge \dots \wedge \Box\Diamond q_n)}_{\varphi_s}$$

where  $p_i, q_j$  are past formulas.

- Add to the game additional variables and a transition relation which encodes the deterministic Buchi automaton
- Examples:  $\Box(p \implies \Diamond q)$ 
  - Introduce a Boolean variable  $x$
  - Initial condition:  $x = 1$
  - Transition relation for the environment:  $\rho_e \wedge (x' = (q \vee x \wedge \neg p))$
  - Winning condition:  $\Box\Diamond x$