

Safety First: A Two-Stage Algorithm for LTL Games

Saqib Sohail Fabio Somenzi

Department of Electrical and Computer Engineering
University of Colorado at Boulder

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Motivation

- Recently, significant algorithmic advances in the game-theoretic approach to synthesis of reactive systems has renewed interest. Piterman 06, Piterman et al 06, Kupferman et al 06, Chatterjee et al 07, Bloem et al 07 are a few examples.
- Despite challenges in scalability, there is increasing hope that synthesis algorithms may be applied to the design and diagnosis of intricate, safety critical protocols.
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- The focus will be on how to avoid some of these challenges without any compromises.

Outline

- 1 Introduction
- 2 Games
- 3 Two Stage Synthesis
 - The Challenge
 - Algorithm
 - Optimizations
 - Implementation
 - Caveats
- 4 Results
- 5 Conclusions

LTL Synthesis - Pnueli and Rosner (POPL'89)

- Automatically build design from specification
- Input:
 - Set of LTL formulae, e.g. $G(req \rightarrow F ack)$, $G(\neg req \rightarrow X(\neg ack))$
 - Partition of the atomic propositions (input/output signals)
 - Environment controls inputs and system controls outputs
- The set of LTL formulae are converted to a non-terminating game with system as **protagonist** and environment as **antagonist**.
- Output: Automatically created functionally correct finite-state machine from the **winning strategy** of the system.
 - If such strategy doesn't exist then the specification is **unrealizable**.

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- In a naive approach, all **formulae** and **automata** are reduced to one deterministic automaton, whose transition structure provides the game graph.
- The **acceptance condition** is taken as the winning condition.
- This approach suffers from the **high cost** of determinization, which is prohibitive for even moderate-sized automata.
- How to avoid the high costs?

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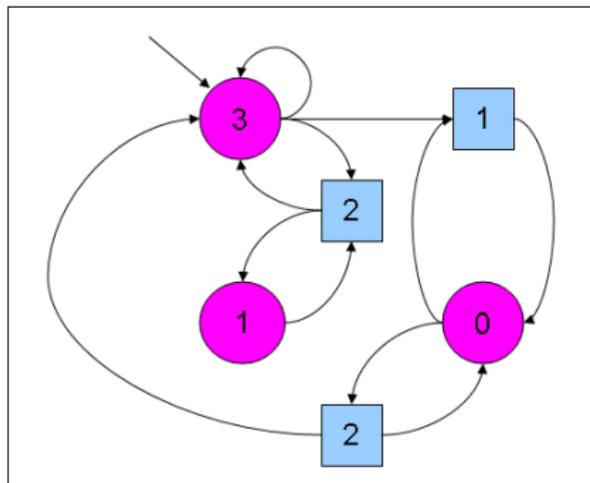
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Example: Game graph with a parity acceptance condition

player0 $\rightarrow \square$
wins if **largest integer**
occurring infinitely often
is **even**

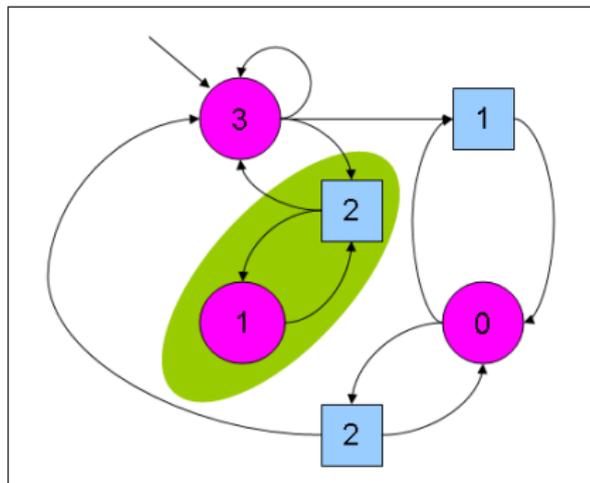
player1 $\rightarrow \bigcirc$
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Game Graphs

- A **game graph** $G = ((S, E), S_0, S_1)$ is a directed graph (S, E) with a finite state space S , a set of edges E and a partition (S_0, S_1) of the state space belonging to player 0 and 1 respectively. We assume that every state has an outgoing edge.
- The game is started by placing a **token** in one of the S_{init} and then this token is moved along the edges, when the token is in a state $s \in S_1$, **player 1** selects one of its outgoing edges and vice-versa. The result is an infinite path in the game graph termed as a **play**.
- A **strategy** for a player is a recipe that specifies how to extend finite path. Formally strategy for **player i** is a function $\sigma : S^* \cdot S_i \rightarrow S$.

Parity Game

- For a game graph $G = (Q, E)$ and a parity function $\pi : Q \rightarrow [k]$, a **parity acceptance condition** requires that the maximal $\pi(s)$ occurring infinitely often is odd (even) for *player1*(0).
- A **generalized parity game** for a game graph $G = (Q, E)$ and a set of parity functions $\{\pi_i | \pi_i : Q \rightarrow [k_i]\}$ is played between the **conjunctive** and **disjunctive** player. The conjunctive player wins if it has a strategy to win all the **parity acceptance conditions** while the disjunctive player wins if it has a strategy for some **parity acceptance condition**.

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Two Game Theoretic Approaches

- The standard approach which is the focus of this talk, requires the determinization of word automata.

$$LTL \rightarrow NBW \rightarrow DRW$$

- The Safraless Approach avoids determinization by working with Tree Automata.

$$LTL \rightarrow NGBW \xrightarrow{\text{realizability}} UGCW \xrightarrow{\text{lang-empt}} UGCT \xrightarrow{\text{optimistic-reduction}} NBT$$

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Specification of a simple 2-Client Arbiter

- Initially there are no acknowledgments.

$$\neg ack_0 \wedge \neg ack_1$$

- The acknowledgments are mutually exclusive.

$$\mathbf{G}(\neg ack_0 \vee \neg ack_1)$$

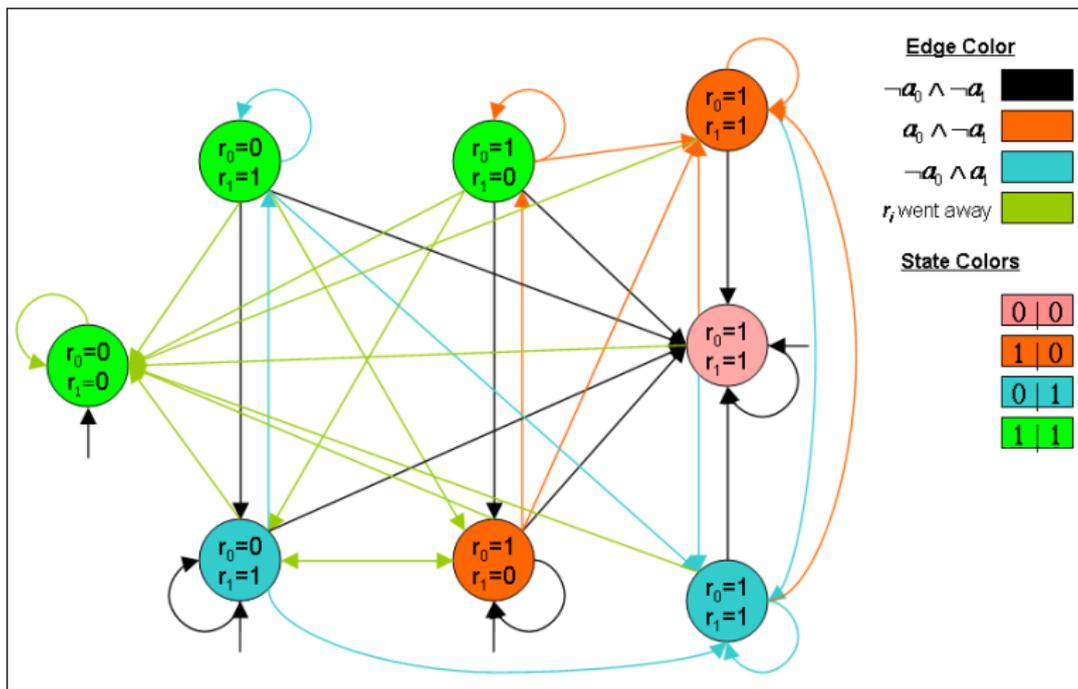
- There are no spurious acknowledgments.

$$\forall i. \mathbf{G}(\neg req_i \rightarrow \mathbf{X}(\neg ack_i))$$

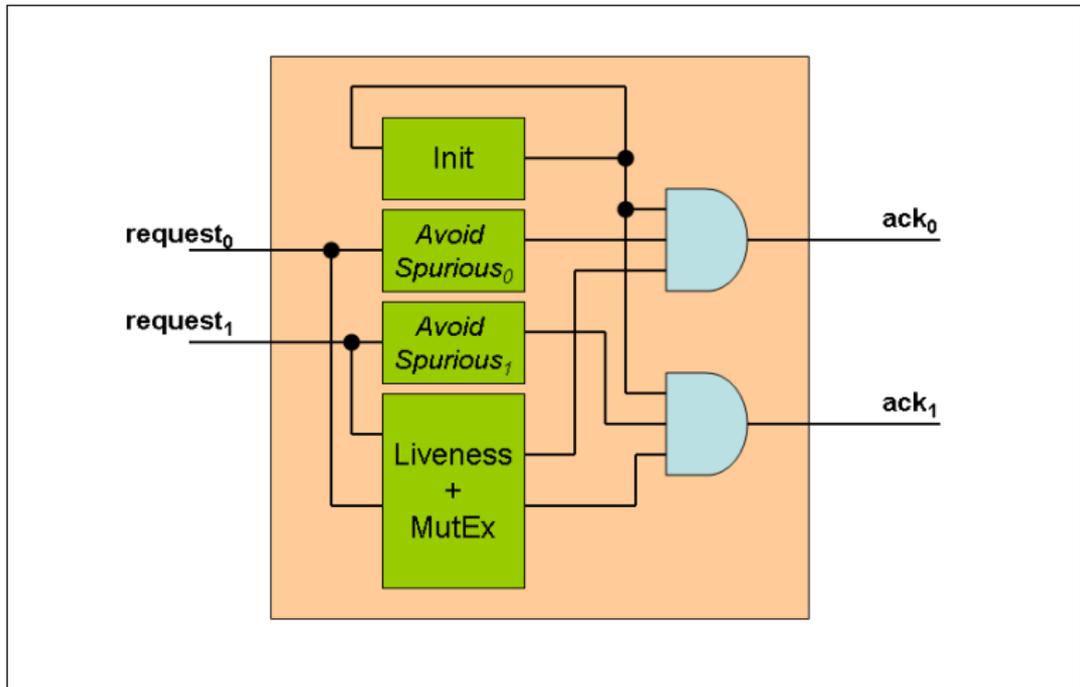
- Every request will eventually be acknowledged

$$\forall i. \mathbf{G}(req_i \rightarrow \mathbf{F} ack_i)$$

Example: Game Graph and Synthesized Strategy

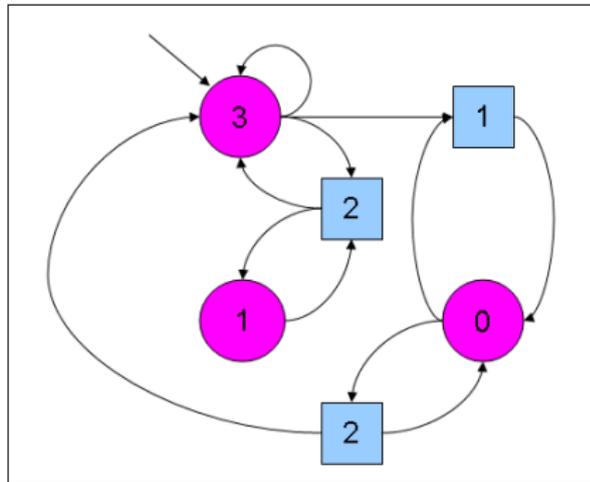


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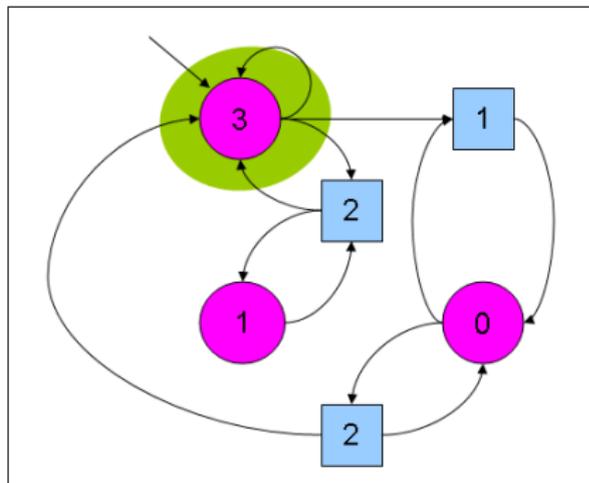
Example: Game play & Strategy Computation for Player 1

○ [Player 1] wins if the maximal $\pi(s)$ occurring infinitely often is **odd**.



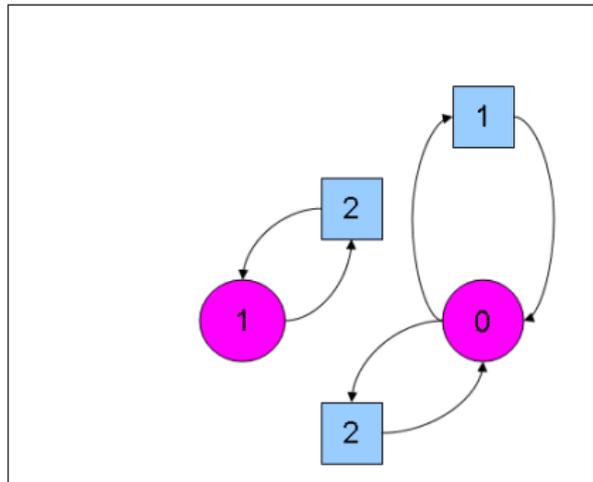
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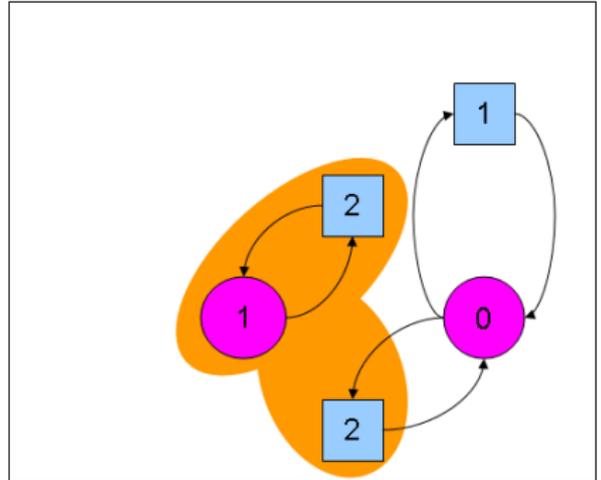
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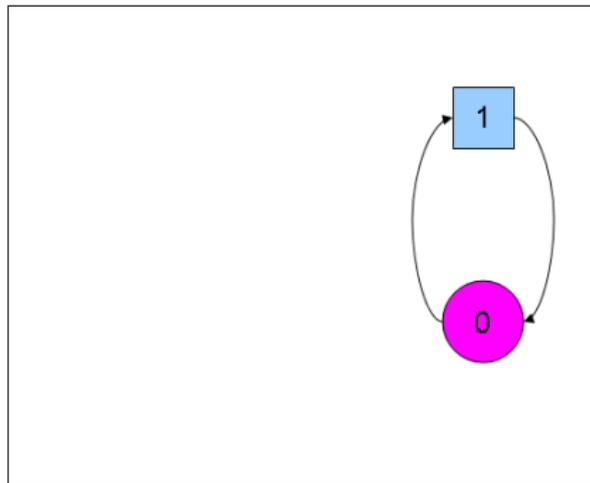
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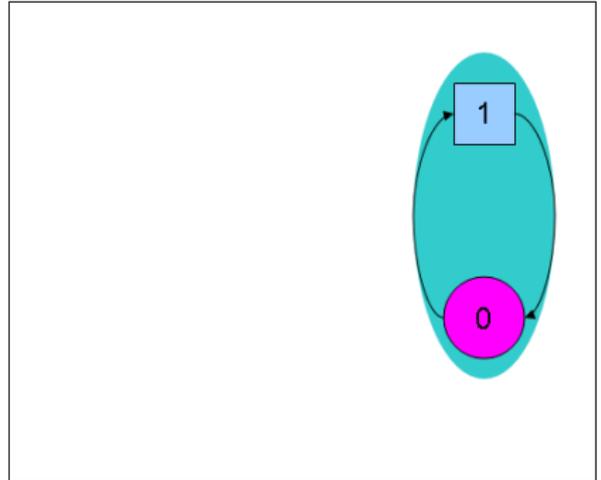
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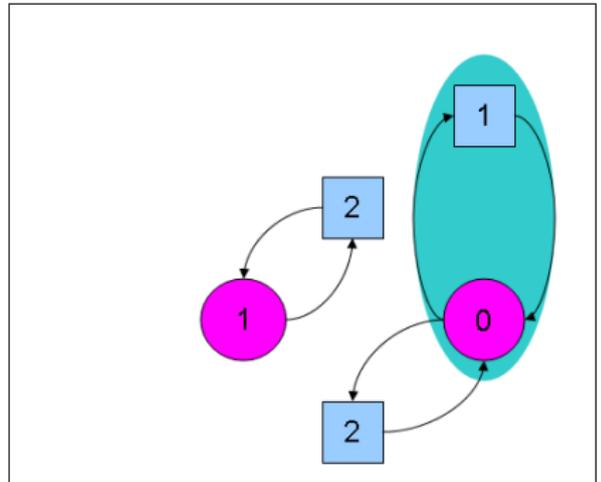
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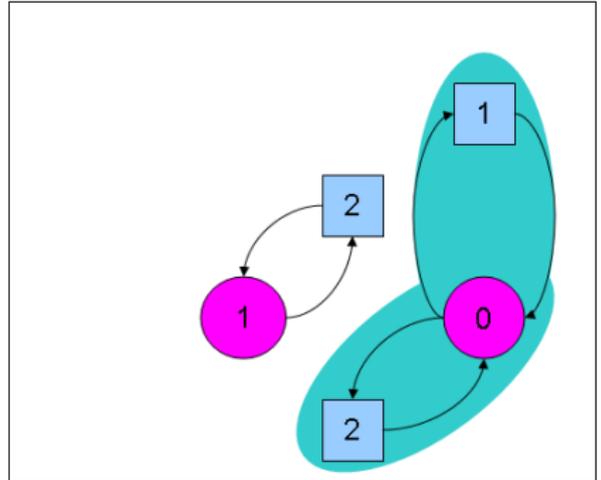
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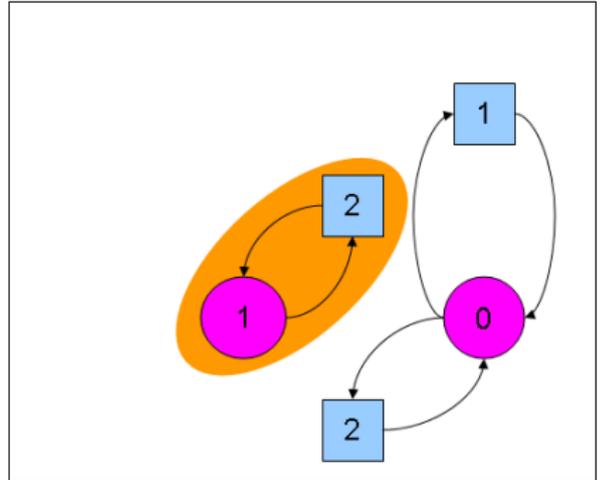
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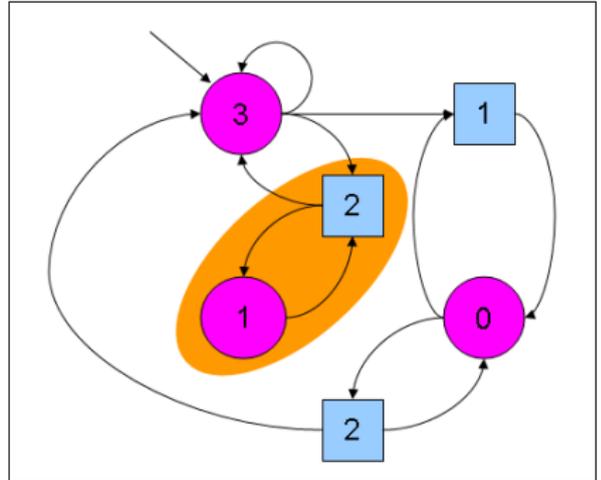
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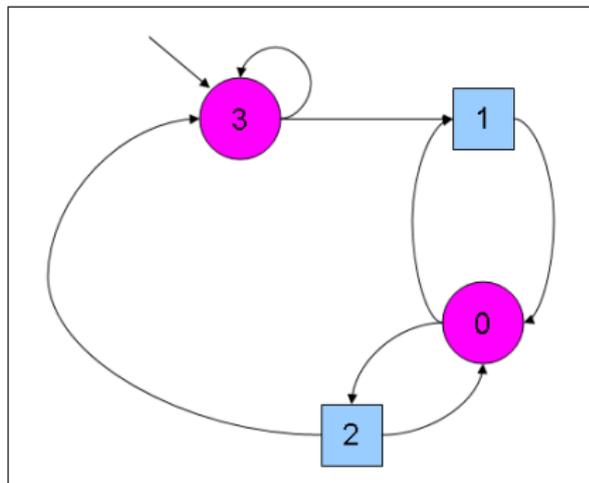
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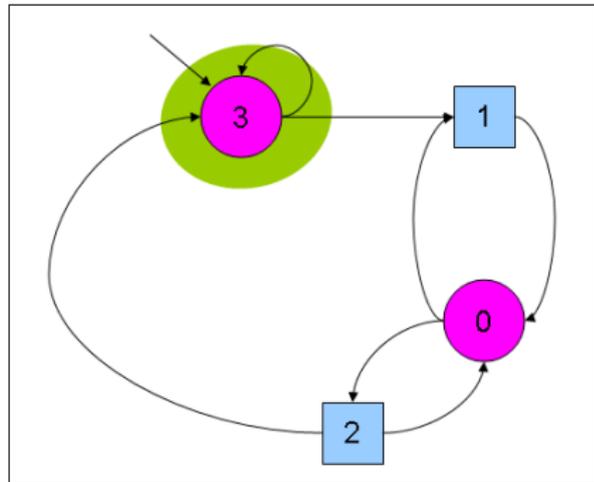
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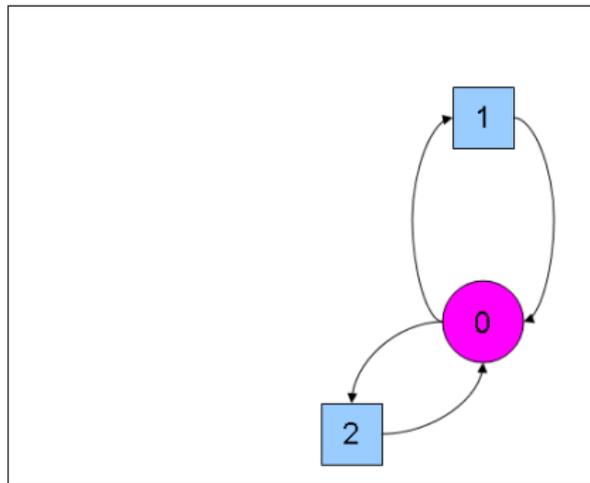
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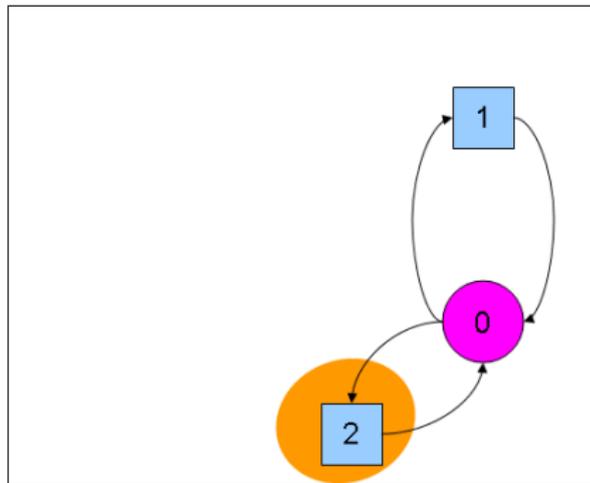
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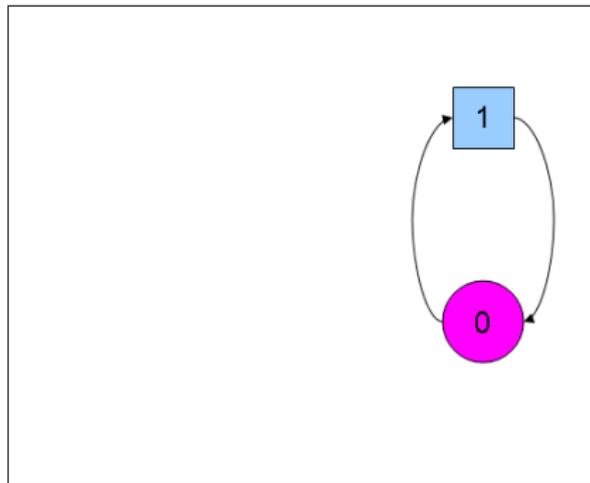
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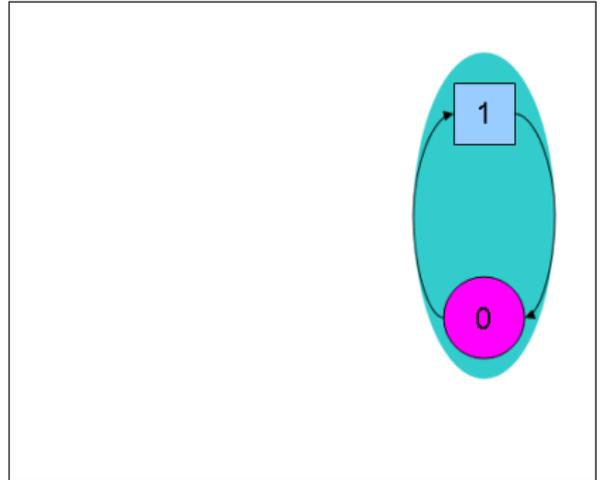
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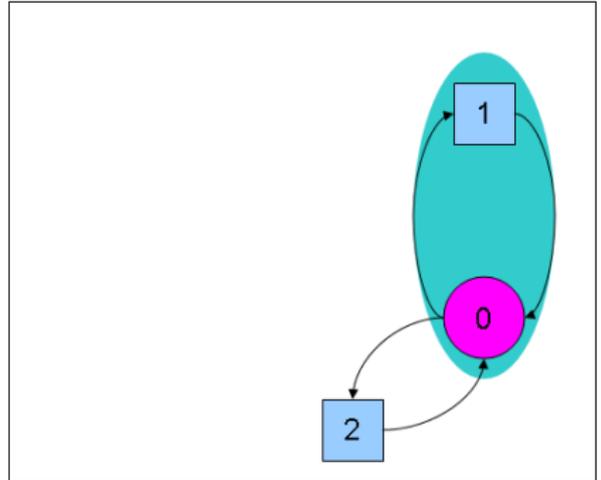
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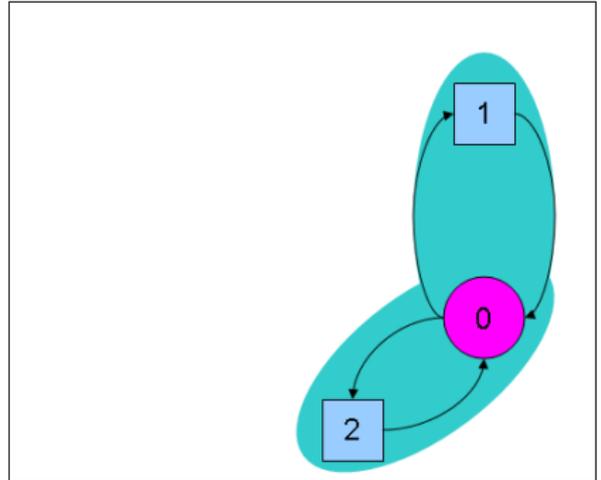
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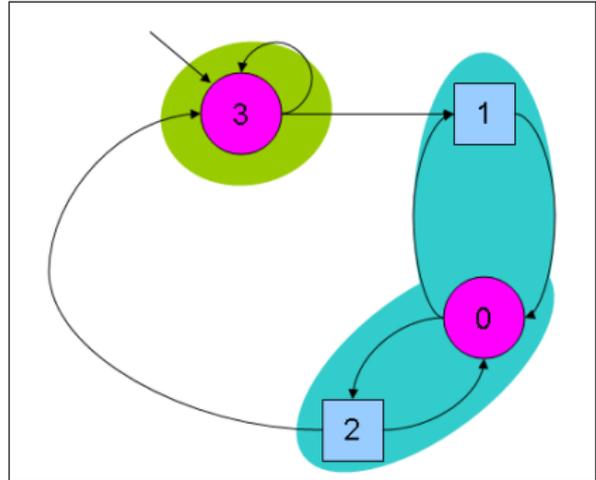
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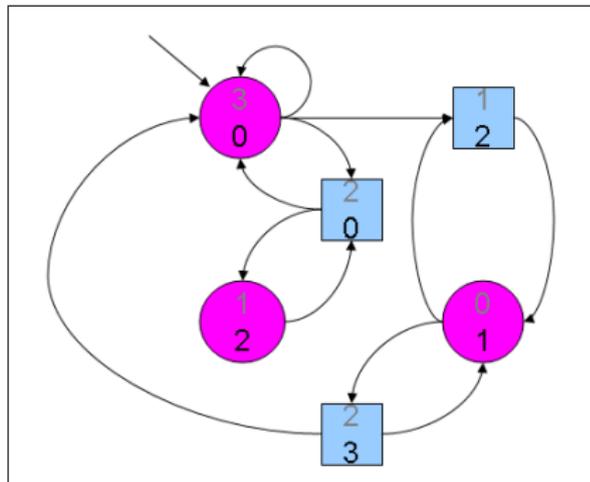
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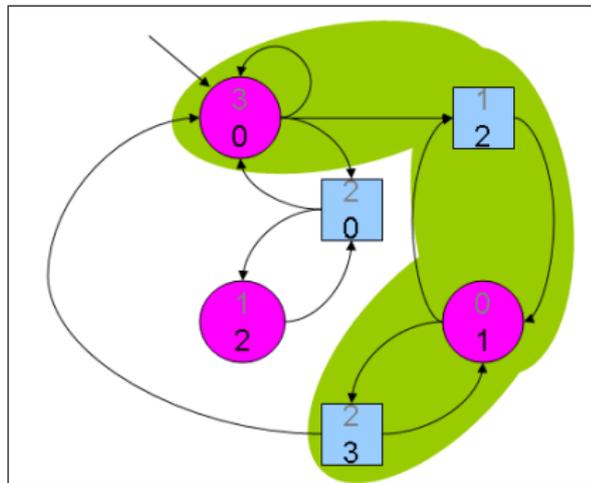
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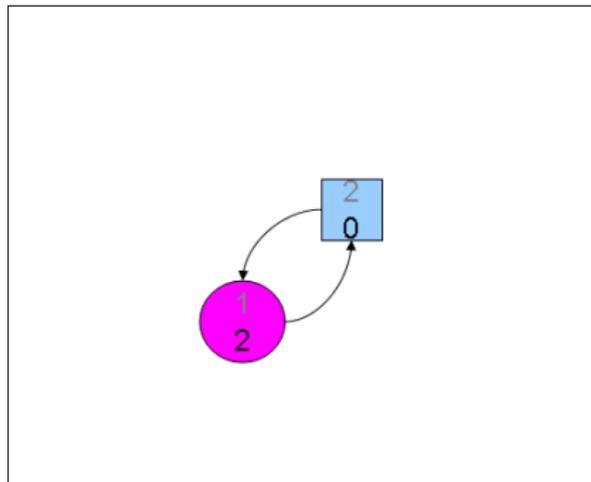
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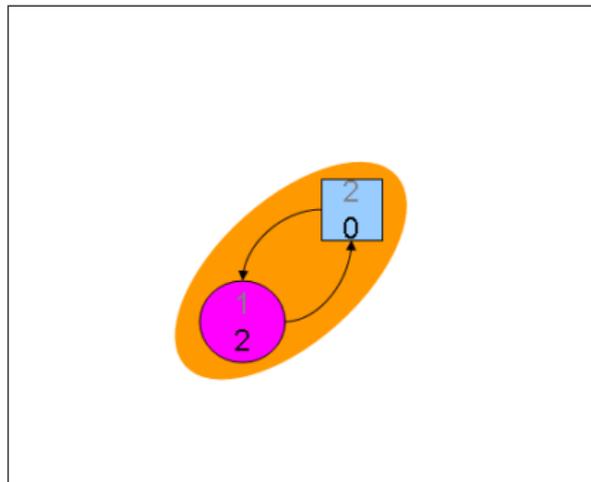
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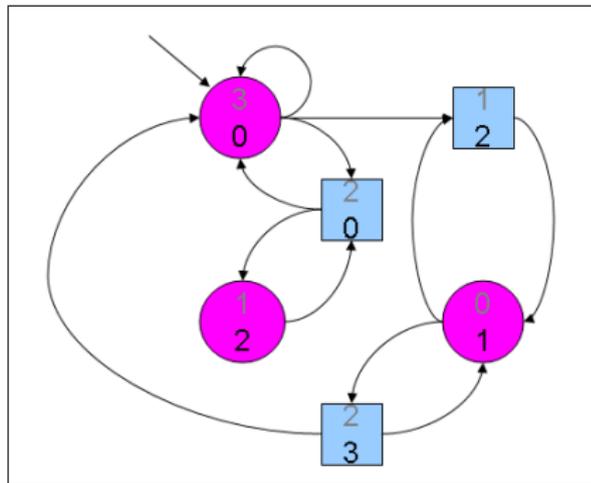
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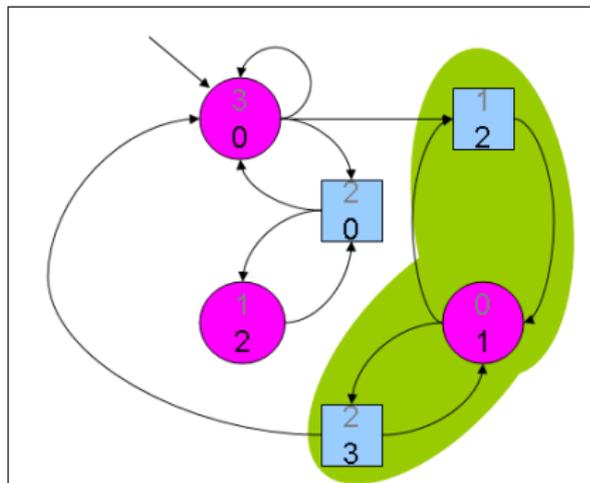
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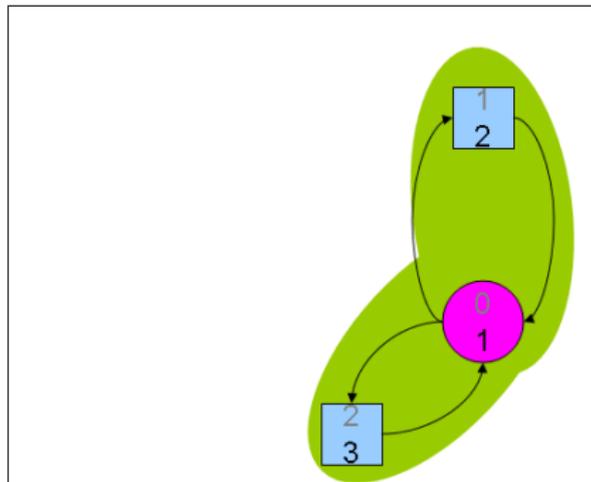
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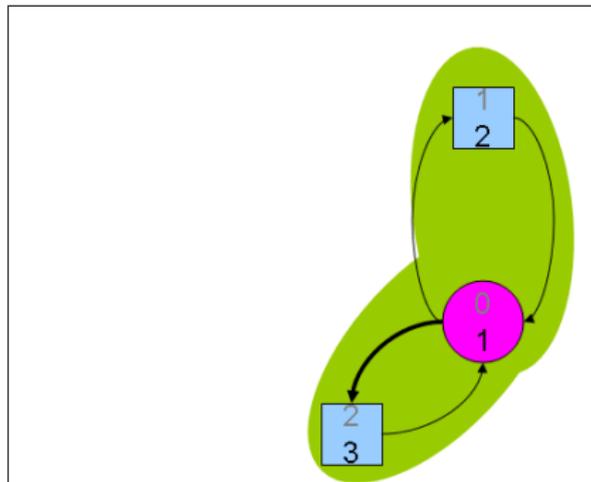
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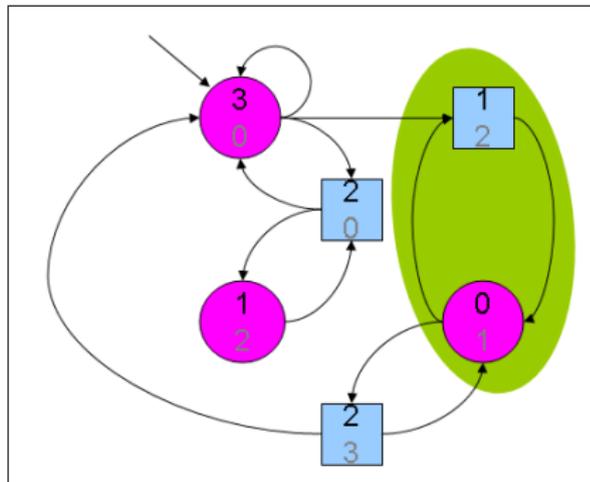
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The Challenge

- Generalized parity game is an **NP-Complete** problem and the current algorithm (Chatterjee et. al 07) is computationally very expensive.
- Is there a simpler solution to the complex problem?
- Is there a way to deal with properties one at a time?

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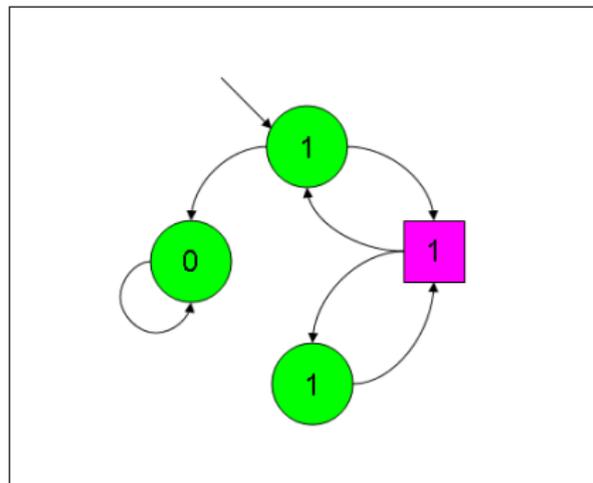
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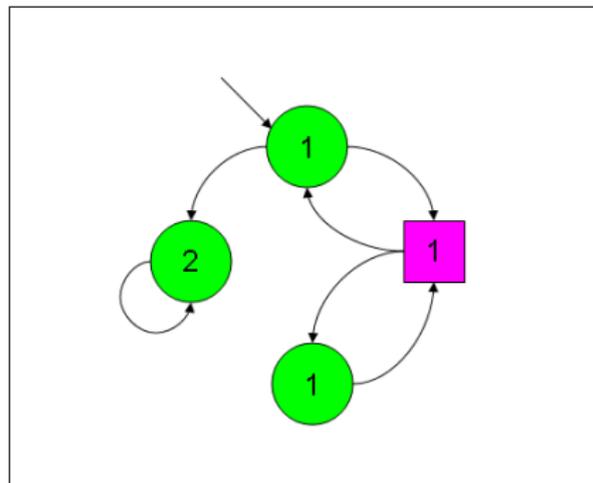
Safety Properties

A **safety condition** for a game graph $G = (Q, E)$ is a function $\pi : Q \rightarrow \{0, 1\}$ such that there is no transition $(u, v) \in E$ such that $\pi(u) = 0$ and $\pi(v) = 1$.



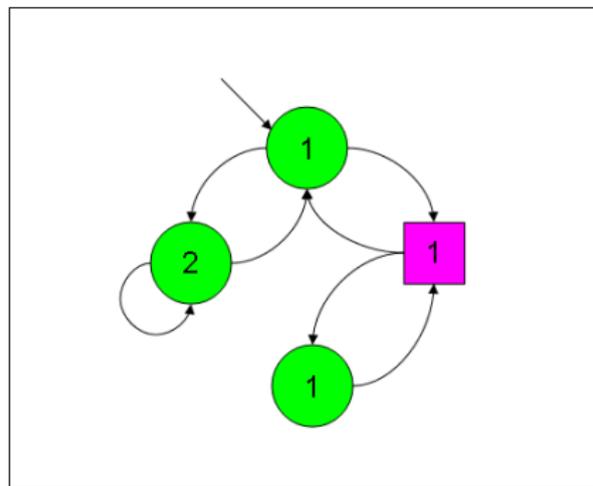
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Persistence Properties

A **persistence condition** for a game graph $G = (Q, E)$ is a function $\pi : Q \rightarrow \{1, 2\}$.



The Claim

- What is so unique about **persistence properties**?
- The winning states for persistence properties can be categorized into **persistent** and **transient** states.
- The computation of strategies is not necessary when we are only interested in determining the persistent and transient states.
- A transient state will stay a transient state for the subsequent games.

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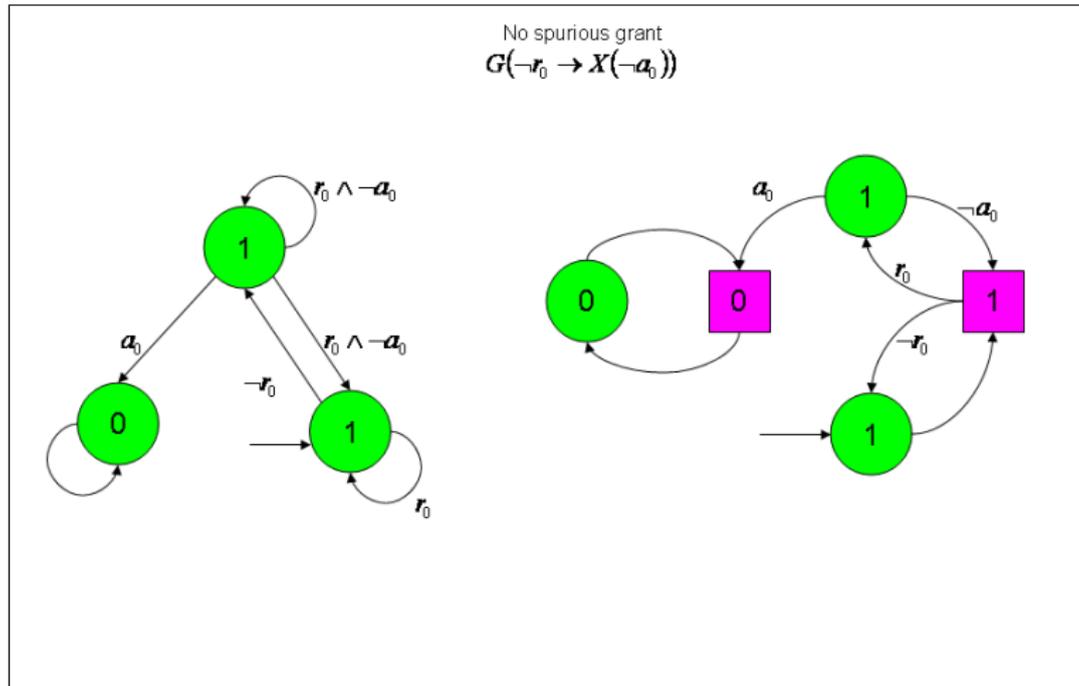
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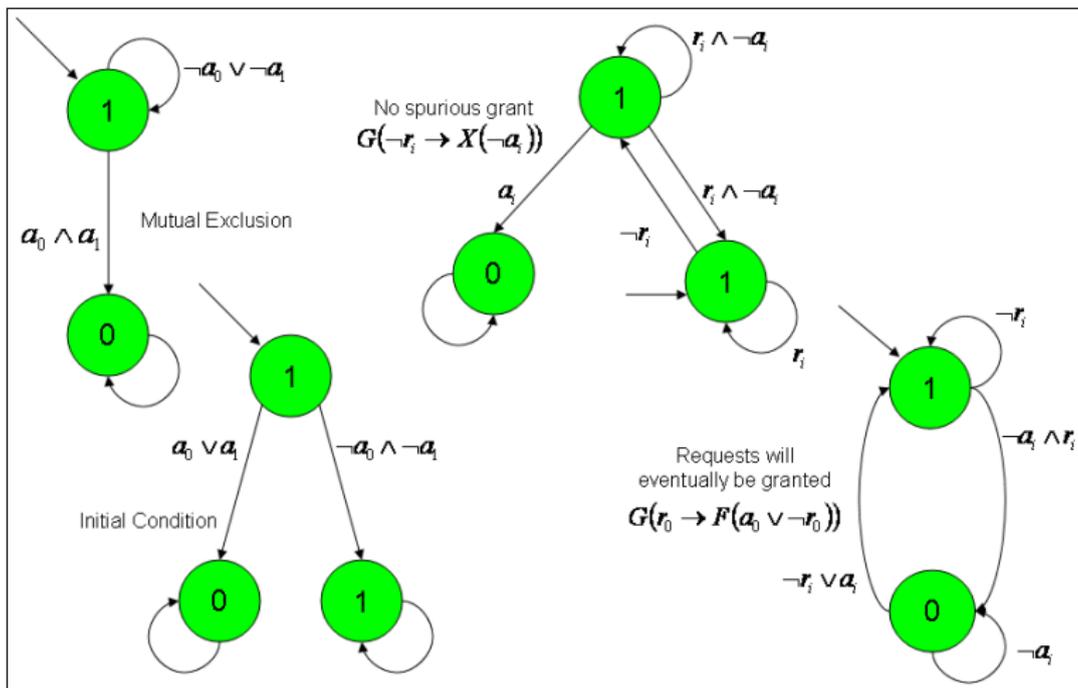
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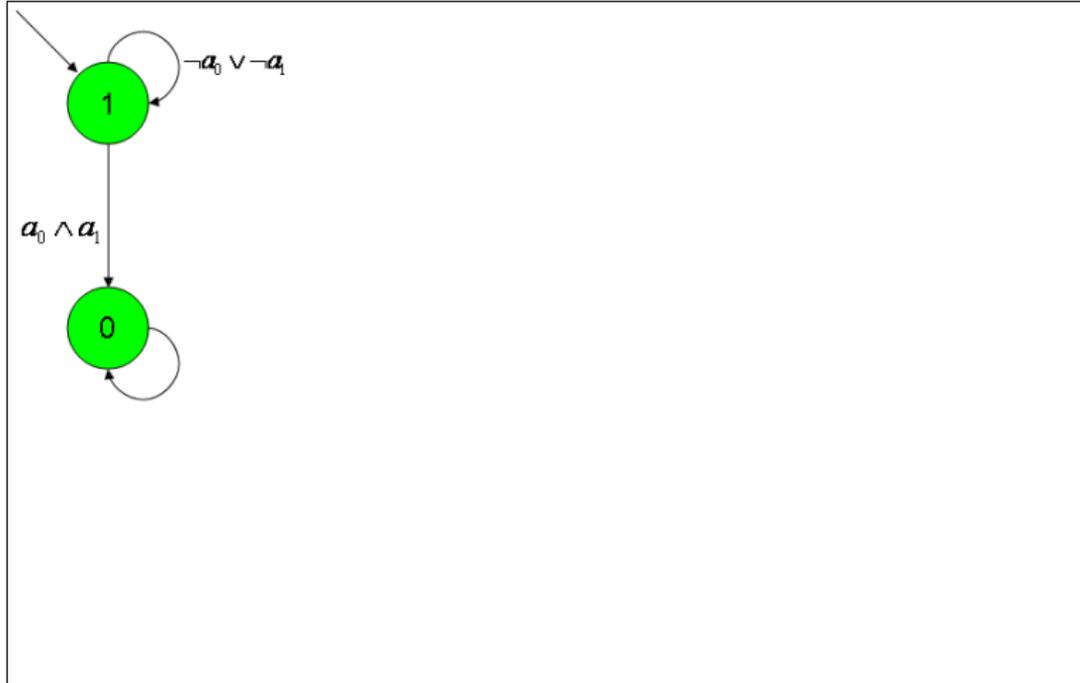
Input/Output based game \rightarrow State based game



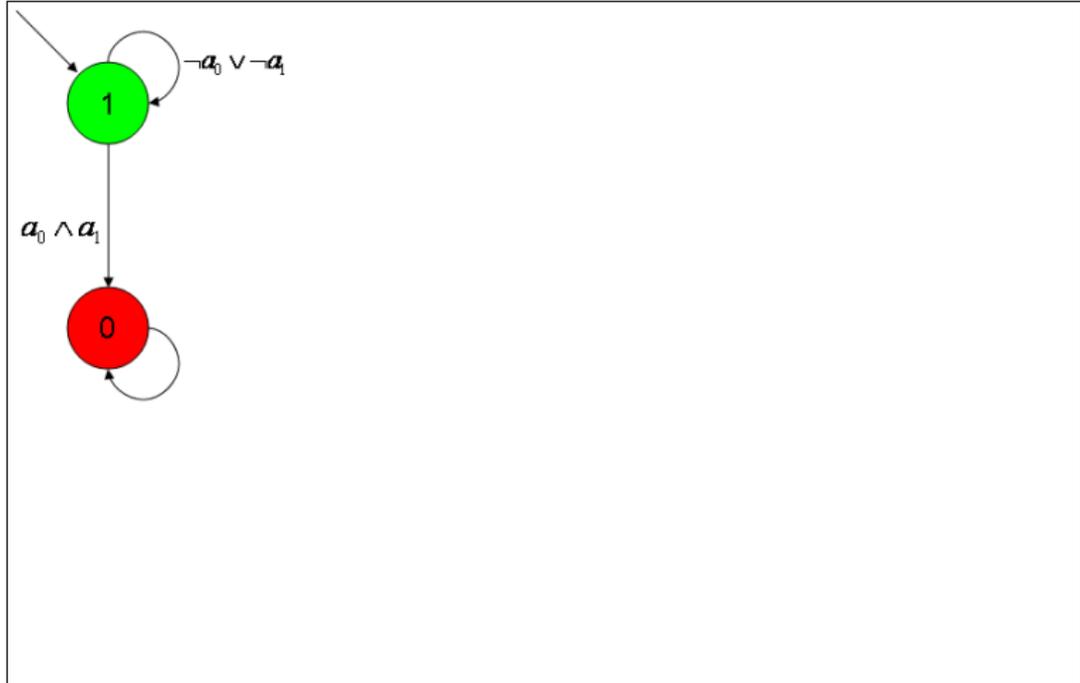
Example: Simple Arbiter revisited



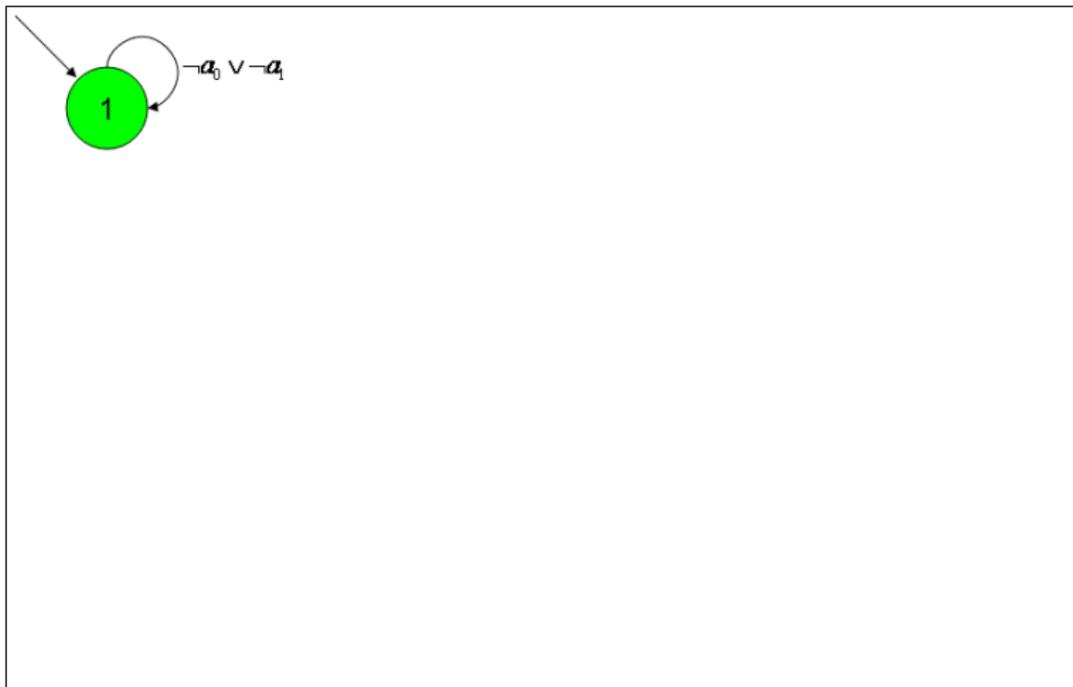
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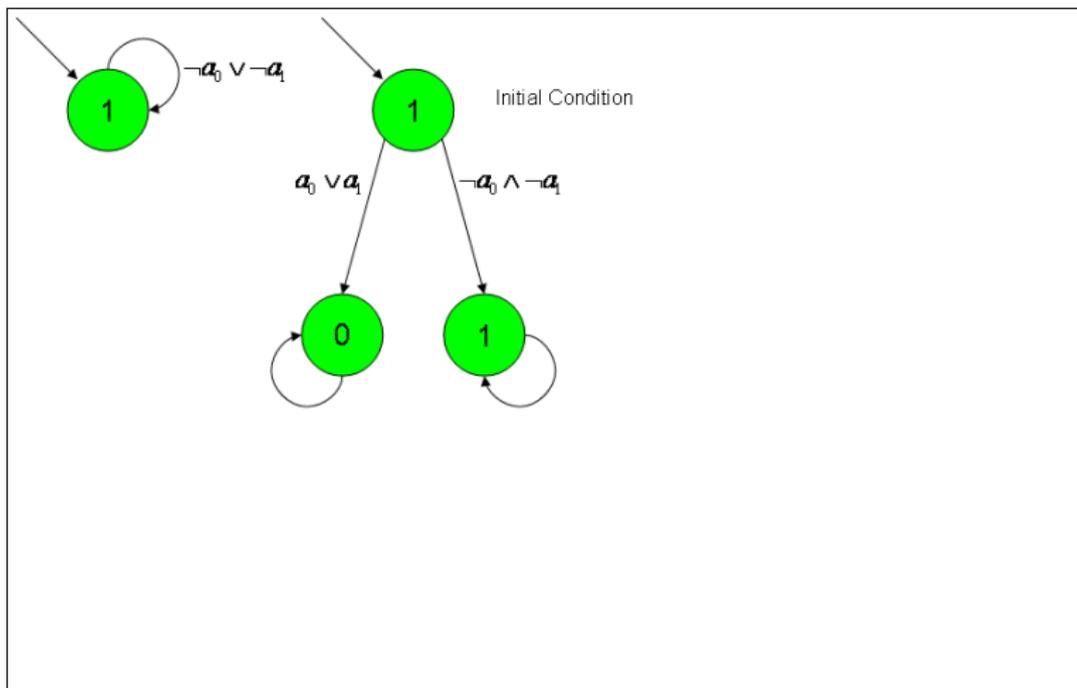
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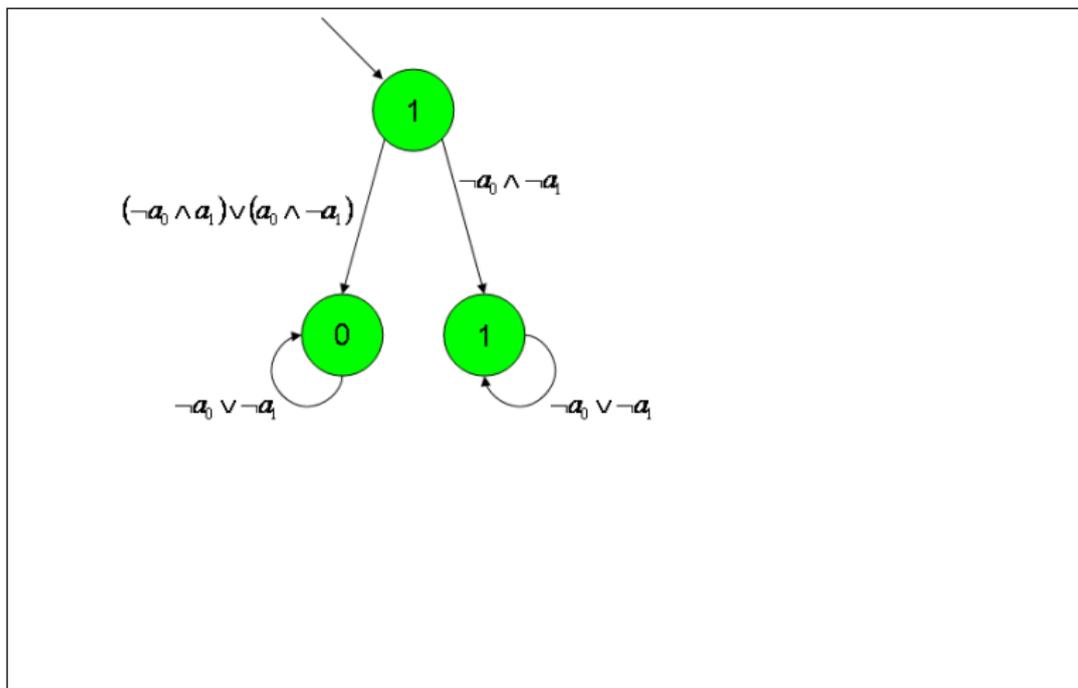
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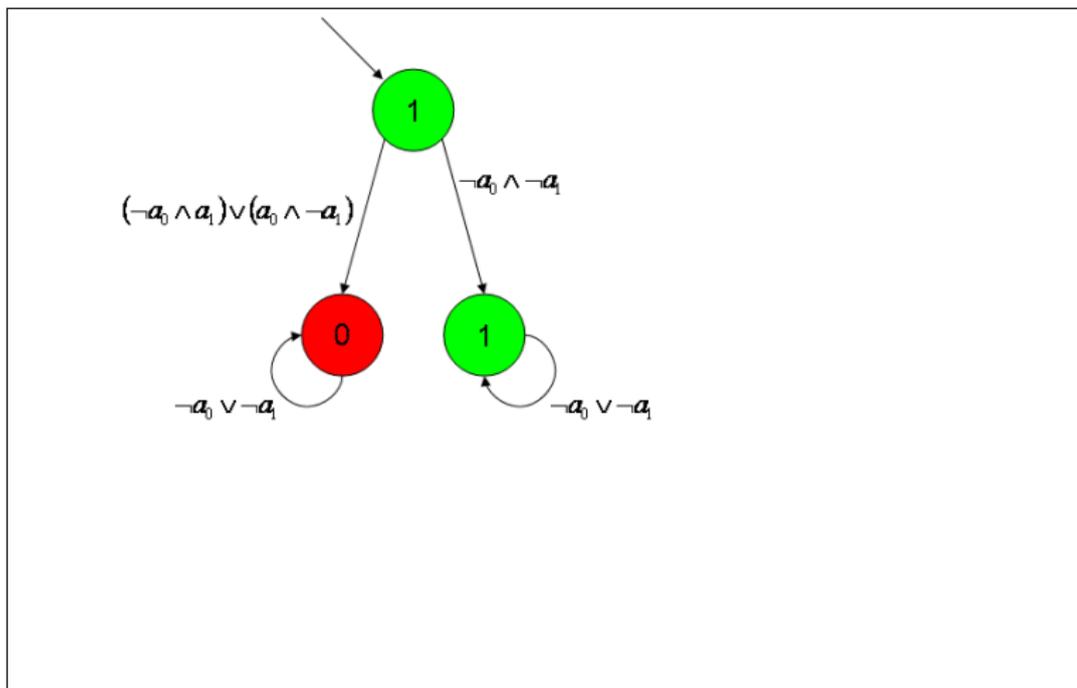
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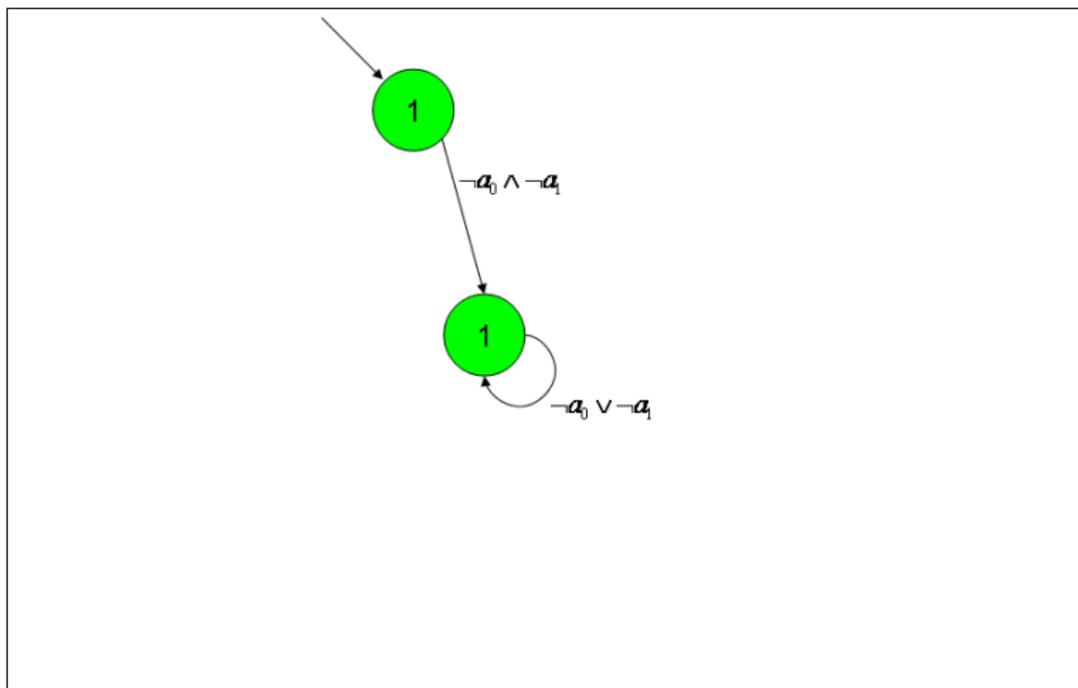
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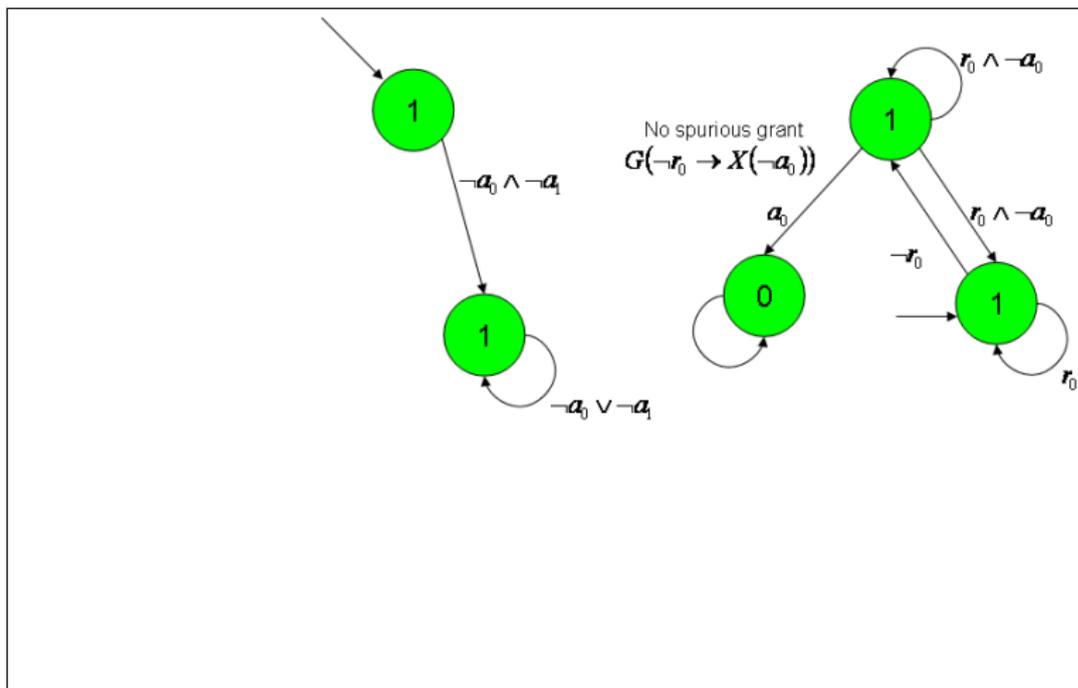
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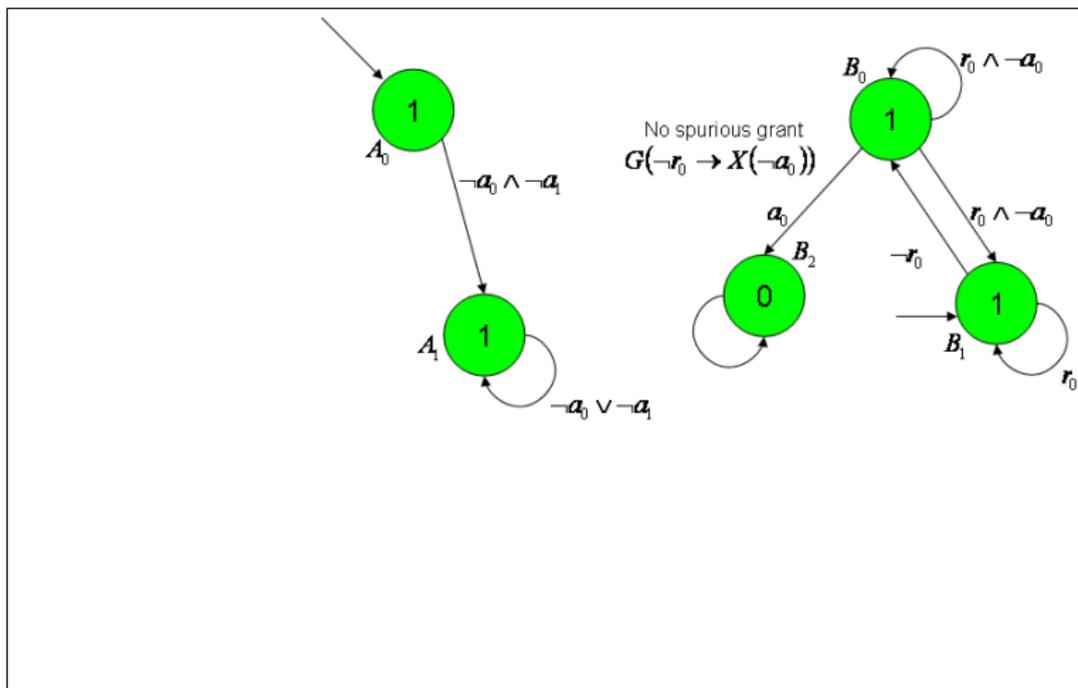
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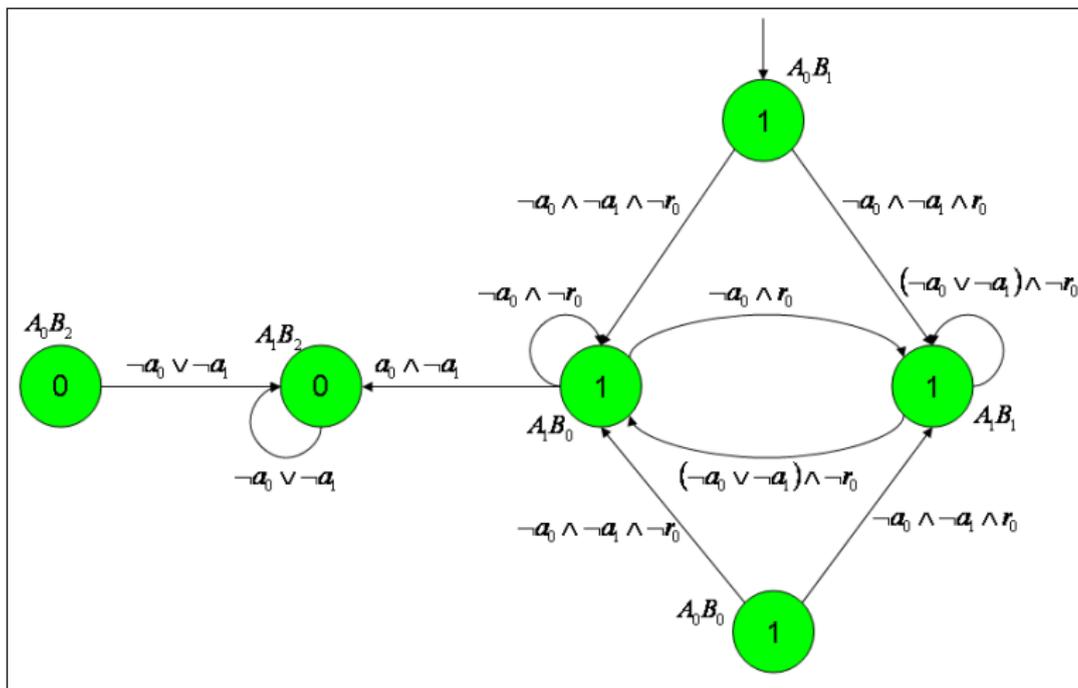
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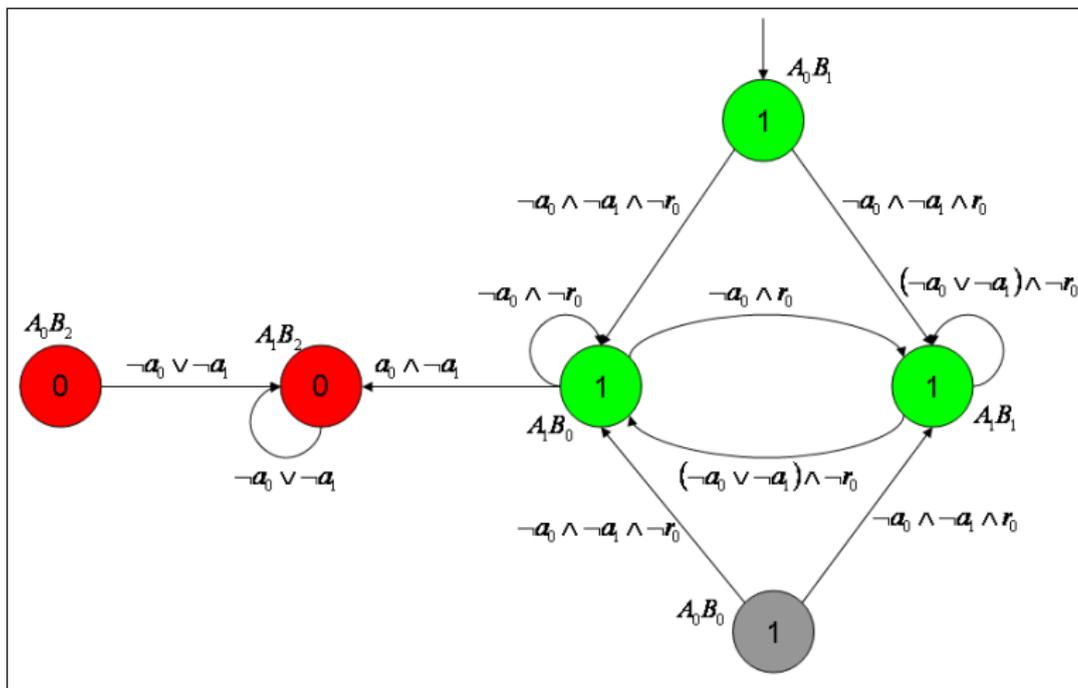
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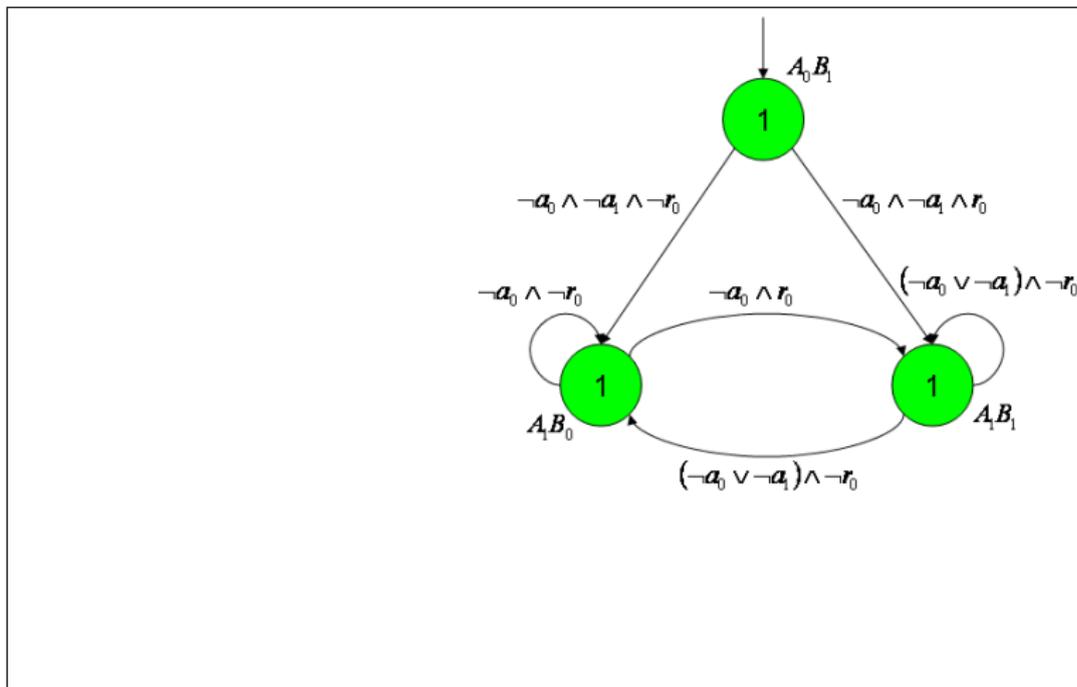
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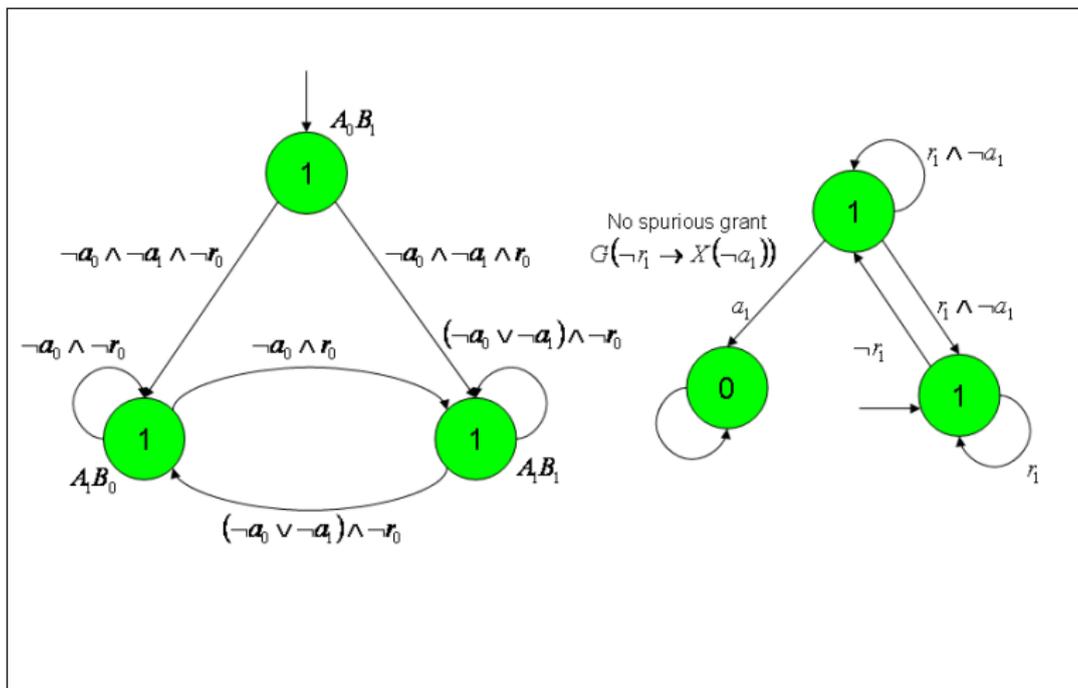
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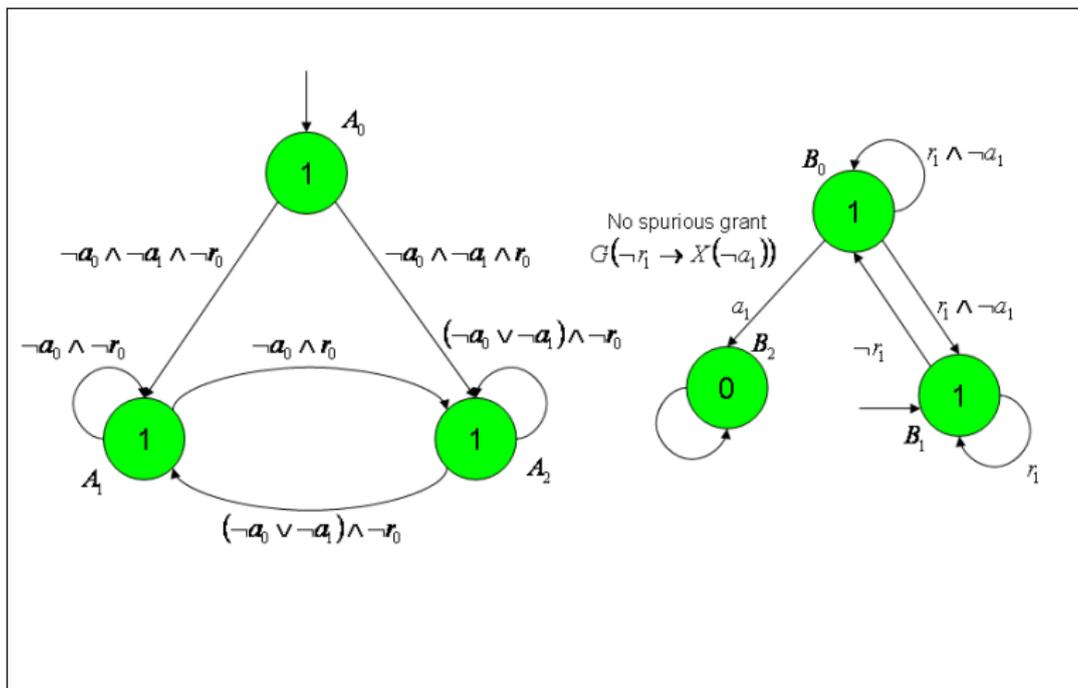
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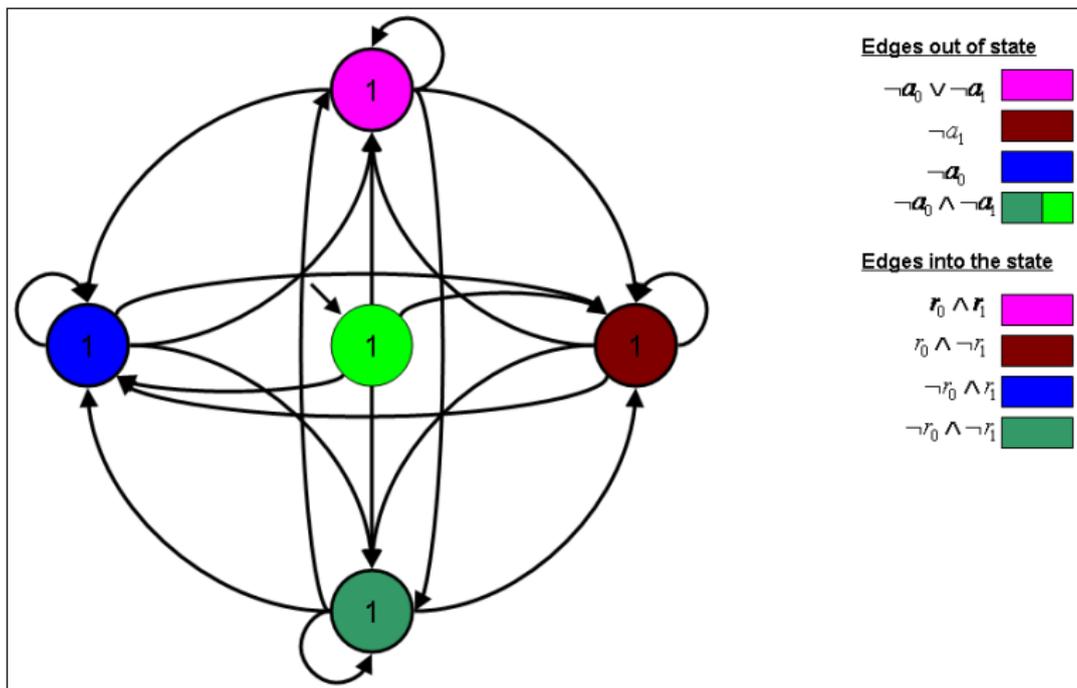
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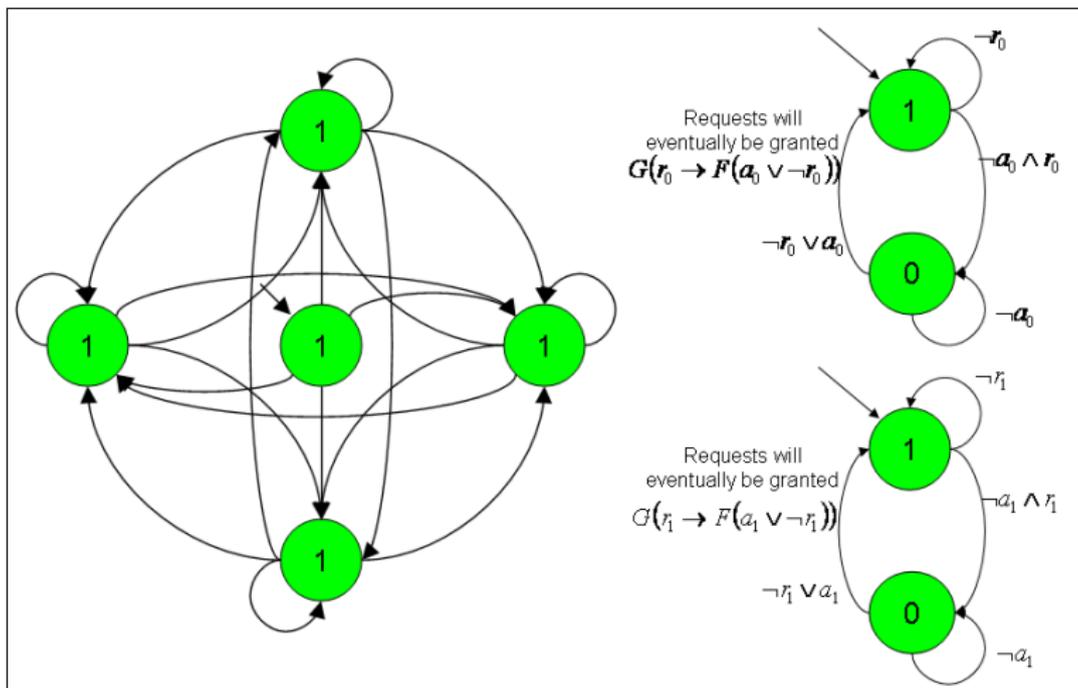
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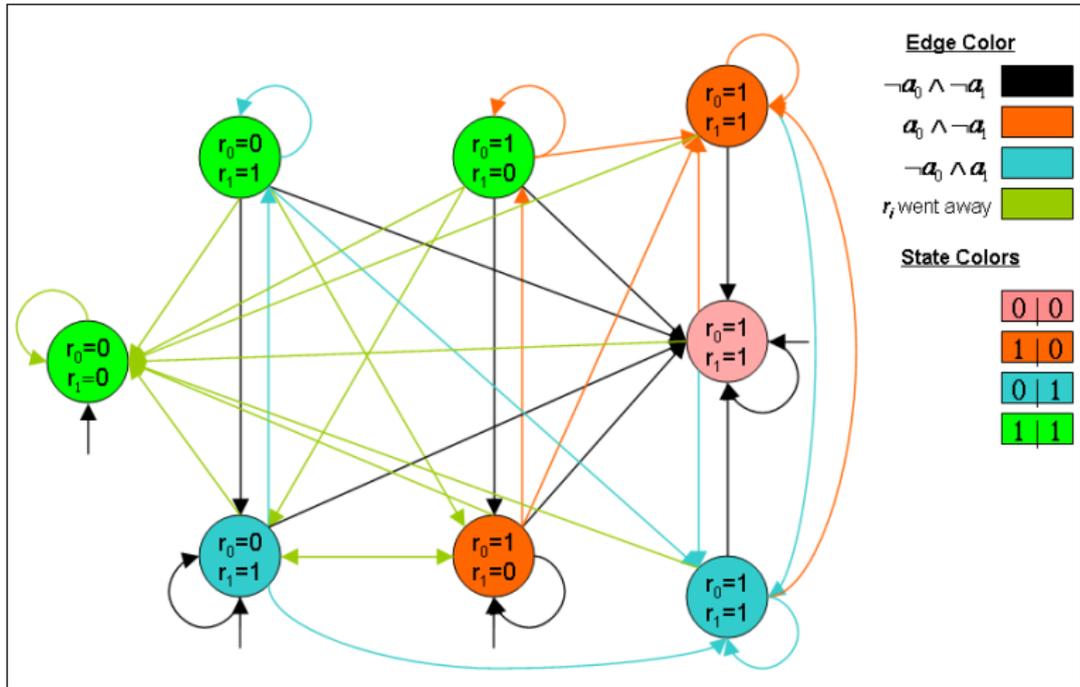
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How significant is the improvement?

- The complexity of “classical” algorithm of (Chatterjee et al 07) is given by

$$O(m \cdot n^{2d}) \cdot \binom{d}{d_1, d_2, \dots, d_k},$$
$$d_i = \lceil k_i/2 \rceil$$

If π_k is a safety condition, solving the game in two stages leads to a better bound for the second stage, $O(m \cdot n^{2d-2}) \cdot \binom{d-1}{d_1, \dots, d_{k-1}}$, while the first stage runs in $O(m \cdot n^2)$.

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Methodology

- Identify the **safety/persistent** properties in the specification.
- Translate each property into a **deterministic** automaton.
- Compose the automaton with already existing game-graph and then playing the 2-player game on the **relevant section** of the graph.
- Determinize all the remaining **non-safety/non-persistent** properties and then compose with the game-graph and play the final generalized parity game on the **relevant section** of the graph.
- Select an appropriate **strategy** which in conjunction with the **property automata** can be translated into **software/hardware**.

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Algorithm

1. **SAFETY-FIRST**($G, SPECIFICATION$)
2. ($SAFETY, NON - PERSISTENT$) $\leftarrow SPECIFICATION$
3. **foreach** $\varphi \in SAFETY$
 - 3.1 $G = G \parallel \text{automaton}_{det}^{\varphi}$
 - 3.2 $(Q_{sys}, E_{sys}) \leftarrow \text{CHATTERJEE}(G, \varphi)$
 - 3.3 $(Q_{new}, E_{new}) \leftarrow \text{OPTIMIZE}(Q_{sys}, E_{sys})$
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5. $(Q_{sys}, E_{sys}, \sigma_{sys}) \leftarrow \text{CHATTERJEE}(G, \varphi_1, \varphi_2, \dots, \varphi_n)$
6. **SYNTHESIZE**($Q_{sys}, E_{sys}, \sigma_{sys}$)

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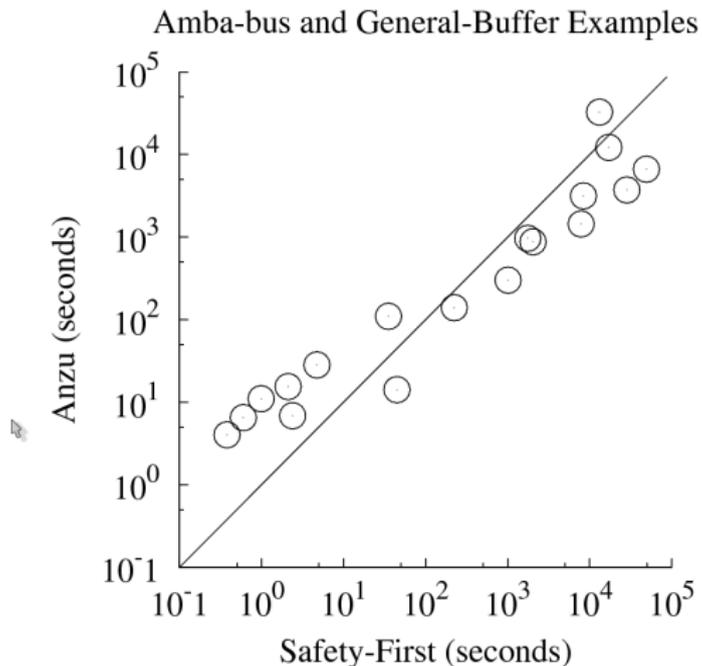
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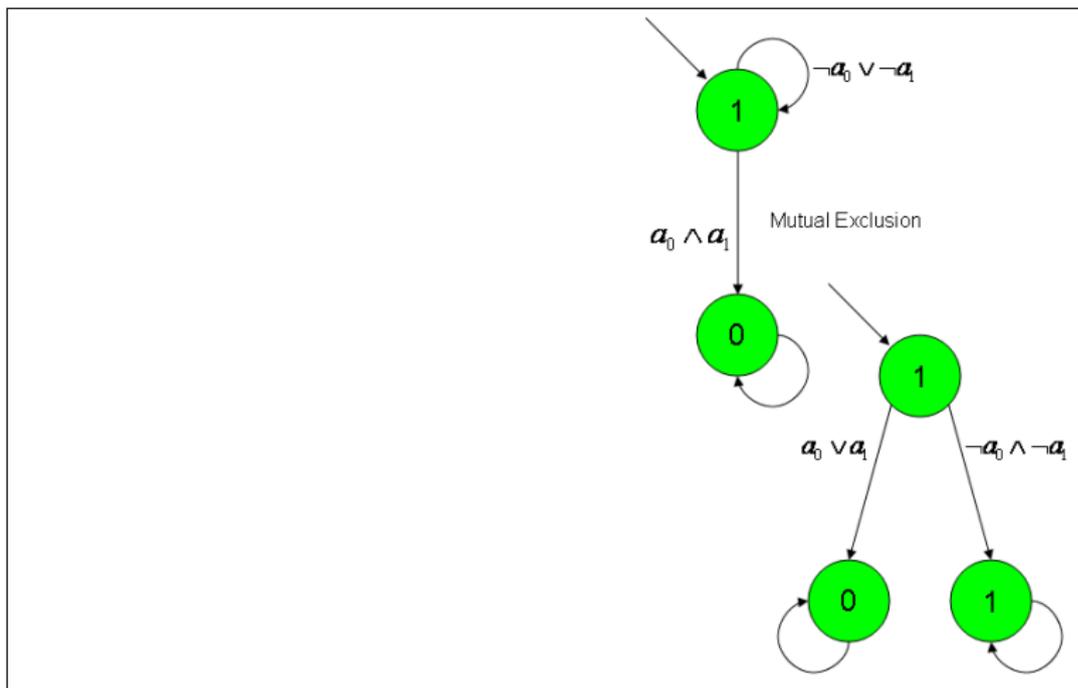
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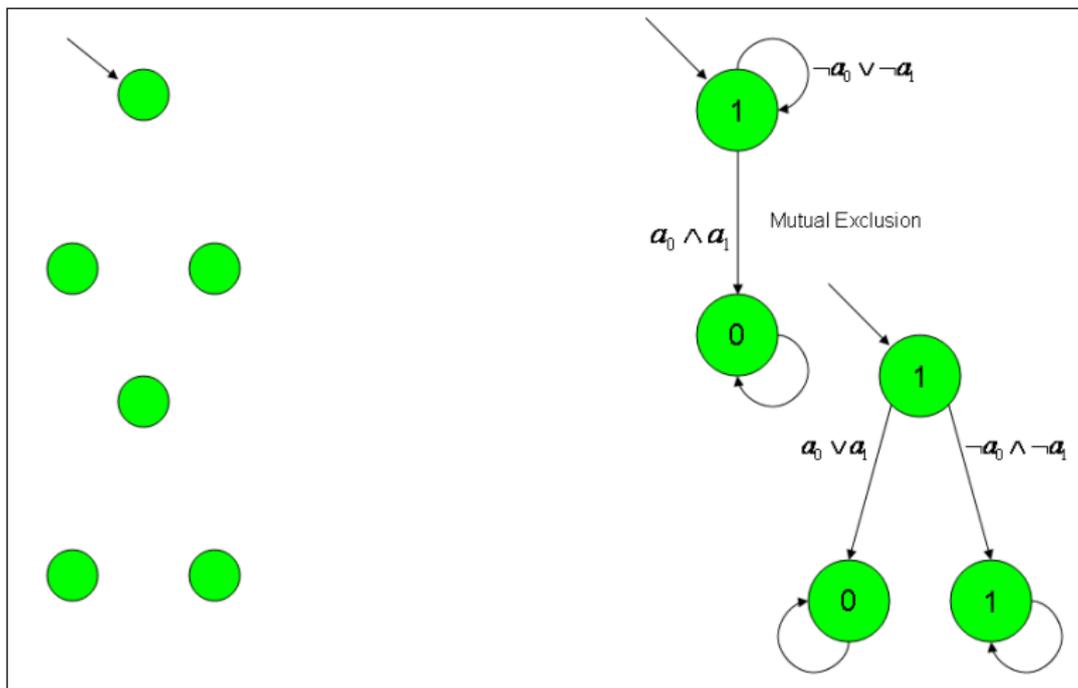
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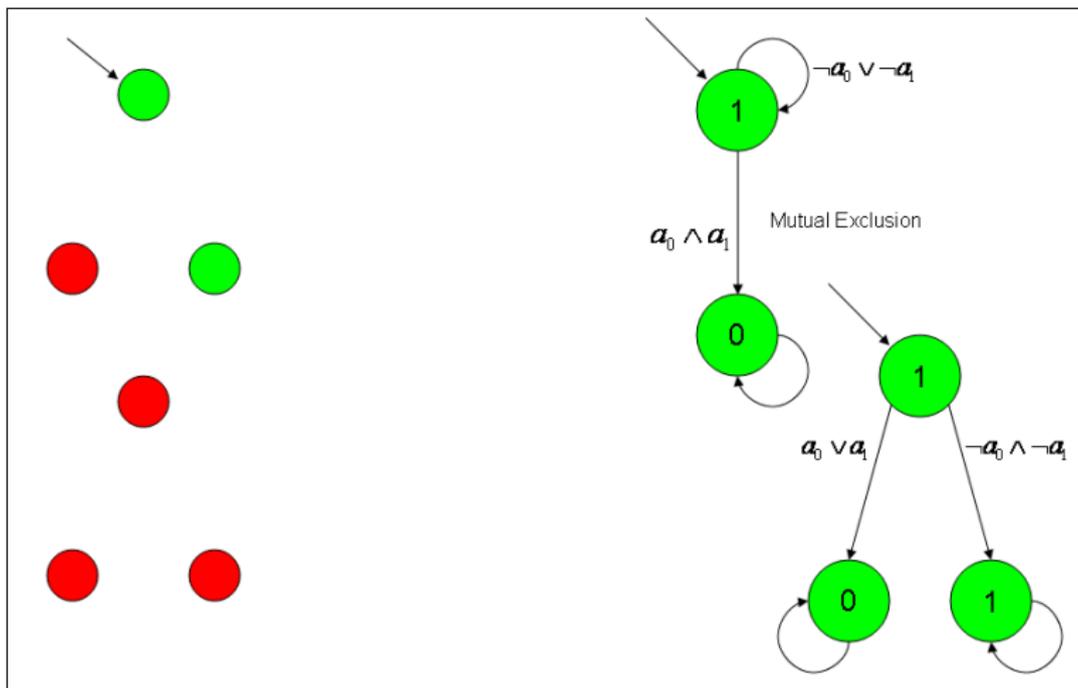
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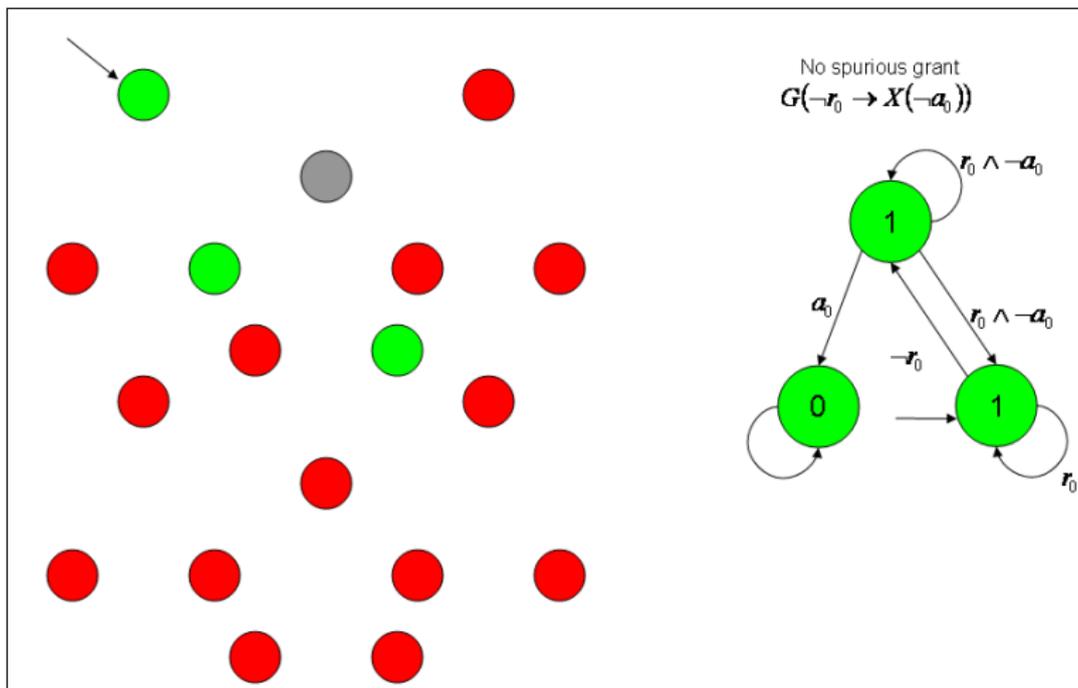
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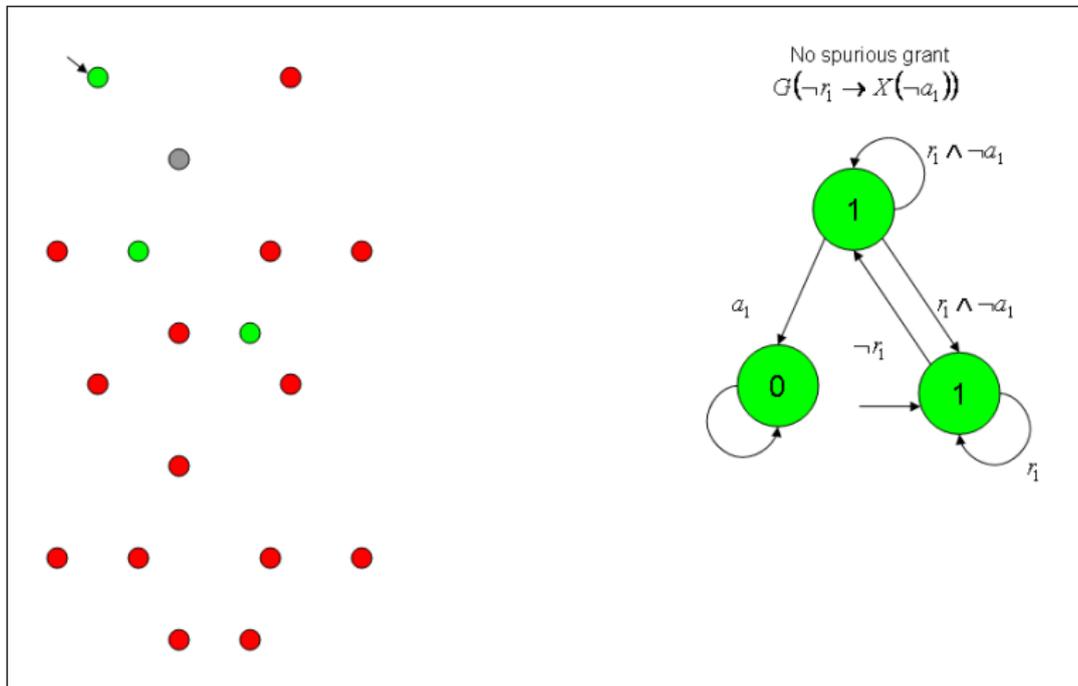
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