# Untitled CAV Paper

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Abstract. This is the abstract

# 1 Introduction

Reactive systems are ubiquitous in real-world problems such as circuit design, industrial automation, or device drivers. Automatic synthesis can provide a *correct by construction* controller for a reactive system from a specification. However, the reactive synthesis problem is 2EXPTIME-complete so naive algorithms are infeasible on even simple systems.

Reactive synthesis is formalised as a game between the *controller* and its *environment*. In this work we focus on safety games, in which the controller must prevent the environment from forcing the game into an error state. Much of the complexity of reactive synthesis stems from tracking the set of states in which a player is winning.

There are several techniques that aim to mitigate this complexity by representing states symbolically. Historically the most successful technique has been to use *Binary Decision Diagrams* (BDDs). BDDs efficiently represent a relation on a set of game variables but in the worst case the representation may be exponential. This means that BDDs are not a one-size-fits-all solution for all reactive synthesis specifications.

Advances in SAT solving technology has prompted research into its applicability to synthesis as an alternative to BDDs. One approach is to find sets of states in CNF [?]. Another approach is to eschew states and focus on *runs* of the game. Previous work has applied this idea to realizability of bounded games [?] by forming abstract representations of the game. In this paper, we extend this idea to unbounded games by constructing approximate sets of winning states from abstract trees.

# 2 Reactive Synthesis

A safety game,  $G = \langle X, L_c, L_u, \delta, I, E, \rangle$ , consists of a set of state variables, sets of controllable and uncontrollable label variables, a transition relationship

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 $\delta: (X, L_c, L_u) \to X$ , an initial state, and an error state. The *controller* and *environment* players choose controllable and uncontrollable labels respectively and the game proceeds according to  $\delta$ .

An run of a game  $(x_0, c_0, u_0), (x_1, c_1, u_1) \dots (x_n, c_n, u_n)$  is a chain of state and label pairs of length n s.t.  $x_{k+1} \leftarrow \delta(x_k, c_k, u_k)$ . A run is winning for the controller if  $x_0 = I \land \forall i \in \{1..n\}(x_i \neq E)$ . In a bounded game of rank n all runs are restricted to length n, whereas unbounded games consider runs of infinite length.

A controller strategy  $\pi^c: X \to L_c$  is a mapping of states to controllable labels. A controller strategy is winning in a bounded game of rank n if all runs  $(x_0, \pi^c(x_0), u_0), (x_1, \pi^c(x_1), u_1) \dots (x_n, \pi^c(x_n), u_n)$  are winning. Bounded realizability is the problem of determining the existence of such a strategy for a bounded game.

An environment strategy  $\pi^e: (X, L_c) \to L_u$  is a mapping of states and controllable labels to uncontrollable labels. A bounded run is winning for the environment if  $x_0 = I \land \exists i \in \{1..n\}(x_i = E)$  and an environment strategy is winning for a bounded game if there exists a run  $(x_0, c_1, \pi^e(x_1, c_1)), (x_1, c_1, \pi^e(x_1, c_1)) \dots (x_n, c_n, \pi^e(x_n, c_n))$  that wins for the environment. Safety games are zero sum, therefore the existence of a controller strategy implies the nonexistence of an environment strategy and vice versa.

#### 2.1 Abstract Game Trees

A set of runs can be symbolically represented by a tree of labels. Each edge of the tree is either a fixed valuation of controllable or uncontrollable variables, or it is unfixed denoting any values are possible. Also, the tree must alternate between controllable and uncontrollable edges to reflect the alternating turns of the game. The entire set of runs of a bounded game of rank n is symbolically represented by a tree of depth 2n and width 1 populated by unfixed edges. Reducing the set of runs in a game forms an abstract game, which can be represented symbolically by an abstract game tree.

A strategy is equivalent to the set of all runs with the player's labels obeying the strategy mapping  $\pi$ . Therefore, a strategy can also be represented by a tree. Relaxing the restriction of the strategy mapping allows for a *partial strategy* in which multiple labels are now possible for a single state.

A strategy or partial strategy can also be thought of as an abstract game. A partial strategy for the controller is a restriction only on the controllable labels in the game. So if the environment can not win in the abstract game equivalent to the controller's partial strategy, then all strategies allowable by that partial strategy must be winning.

An abstract game tree can be checked for the existence of a winning run by a SAT solver. The tree must be encoded into CNF by making copies of the transition relation for each game step in tree. The formula must also check whether the error state has been reached in no branch/in all branches for the controller and environment respectively. The functions that produce these formulas are shown in Algorithm 1.

## Algorithm 1 Tree formulas for Controller and Environment respectively

```
function TREEFORMULA(gt)

if RANK(gt) == 0 then

return \neg E(x^{gt})

else

return \neg E(x^{gt}) \land \bigwedge_{n \in SUCC(gt)} (\delta(n) \land LABEL(n) \land TREEFORMULA(n))

end if

end function

function \overline{treeFormula}(gt)

if RANK(gt) == 0 then

return E(x^{gt})

else

return E(x^{gt})

else

return E(x^{gt}) \lor \bigvee_{n \in SUCC(gt)} (\delta(n) \land LABEL(n) \land \overline{treeFormula}(n))

end if

end function
```

When we produce a formula using TREEFORMULA the SAT solver is playing on behalf on the controller. Any unfixed labels in the tree, controllable or uncontrollable, will be existentially quantified. This means that if there exists a way for the game to be solved when both players cooperate the SAT solver will find it. If no winning run exists in an abstract game even when the players are cooperating then there is definitely no winning run when the opponent is playing adversarily.

### 2.2 Counterexample Guided Bounded Synthesis

By checking for the existence of a winning run for the environment in an abstract game tree constructed from a partial controller strategy we are checking if the partial strategy is winning. If no spoiling run can be found then the partial strategy must always win. If a spoiling run is found then we then we have a counterexample that proves that the controller strategy is not winning. This forms the basis of a counterexample guided abstraction refinement framework that operates on candidate strategies.

Synthesis begins with an empty game as seen in Figure 1a. Initially we are playing on behalf of the environment because it chooses the first move in each step. The empty game is passed to the SAT solver, which searches for a candidate environment strategy. If a candidate is found (Figure 1b) then it is checked for a spoiling strategy by solving for the controller. If no spoiling strategy exists, that means our candidate is a winning strategy and the algorithm terminates. Otherwise we find a counterexample (Figure 1c), which is used to refine the empty game tree to include the first move from the controller's spoiling strategy (Figure 1d. The algorithm continues by finding a new candidate for the environment (Figure 1e), and a new counterexample to refine the game abstraction once again (Figure 1f).

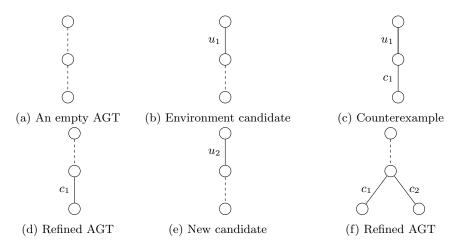


Fig. 1: Abstract game trees  $\,$