

# Towards designing intelligent machines via reactive synthesis

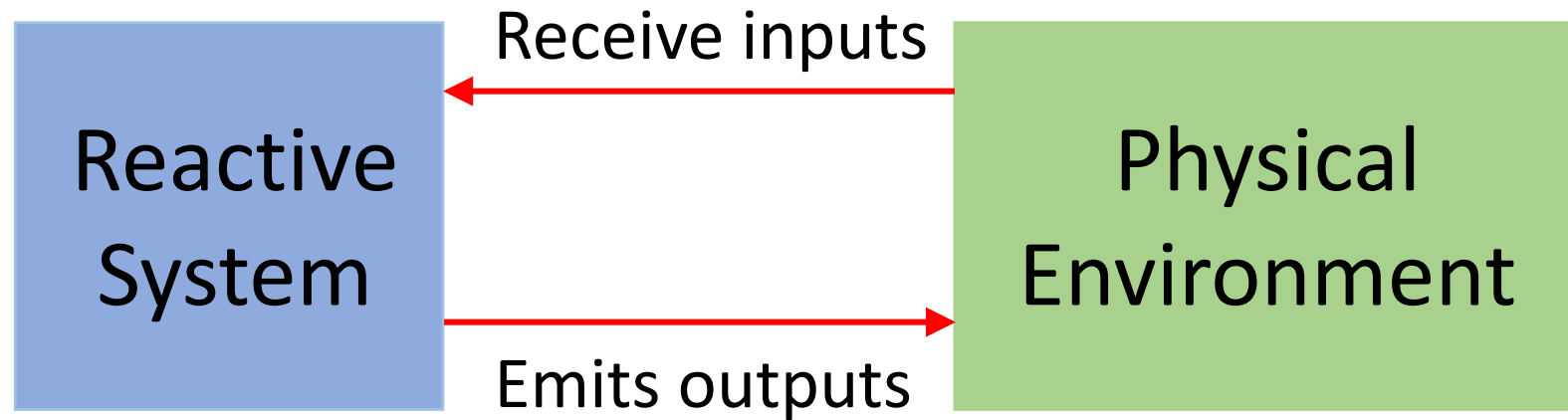
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# Reactive systems



Continuous cycle of interaction

# “Reactive” systems today

## Robot and human interactions



## Autonomous vehicles

AARIAN MARSHALL AND ALEX DAVIES TRANSPORTATION 05.24.18 03:38 PM

## UBER'S SELF-DRIVING CAR SAW THE WOMAN IT KILLED, REPORT SAYS

...The problem is that it's hard to find images of every sort of situation that could happen in the wild. Can the system distinguish a tumbleweed from a toddler. ...

Designing correct reactive systems is hard

**Specifying intent of a reactive program is easier**

Specifying intent of a reactive program is easier

Can we automatically generate a reactive program from its specification?

Reactive synthesis

Towards designing  
intelligent machines  
via  
reactive synthesis

Richness  
Specifications

Formal  
Guarantees

Scalability

# On today's agenda

- Reactive synthesis for planning
  - Qualitative and Quantitative specifications
- Automata-based quantitative reasoning
- Generality of approach in Formal Quantitative Reasoning
  - Beyond reactive synthesis

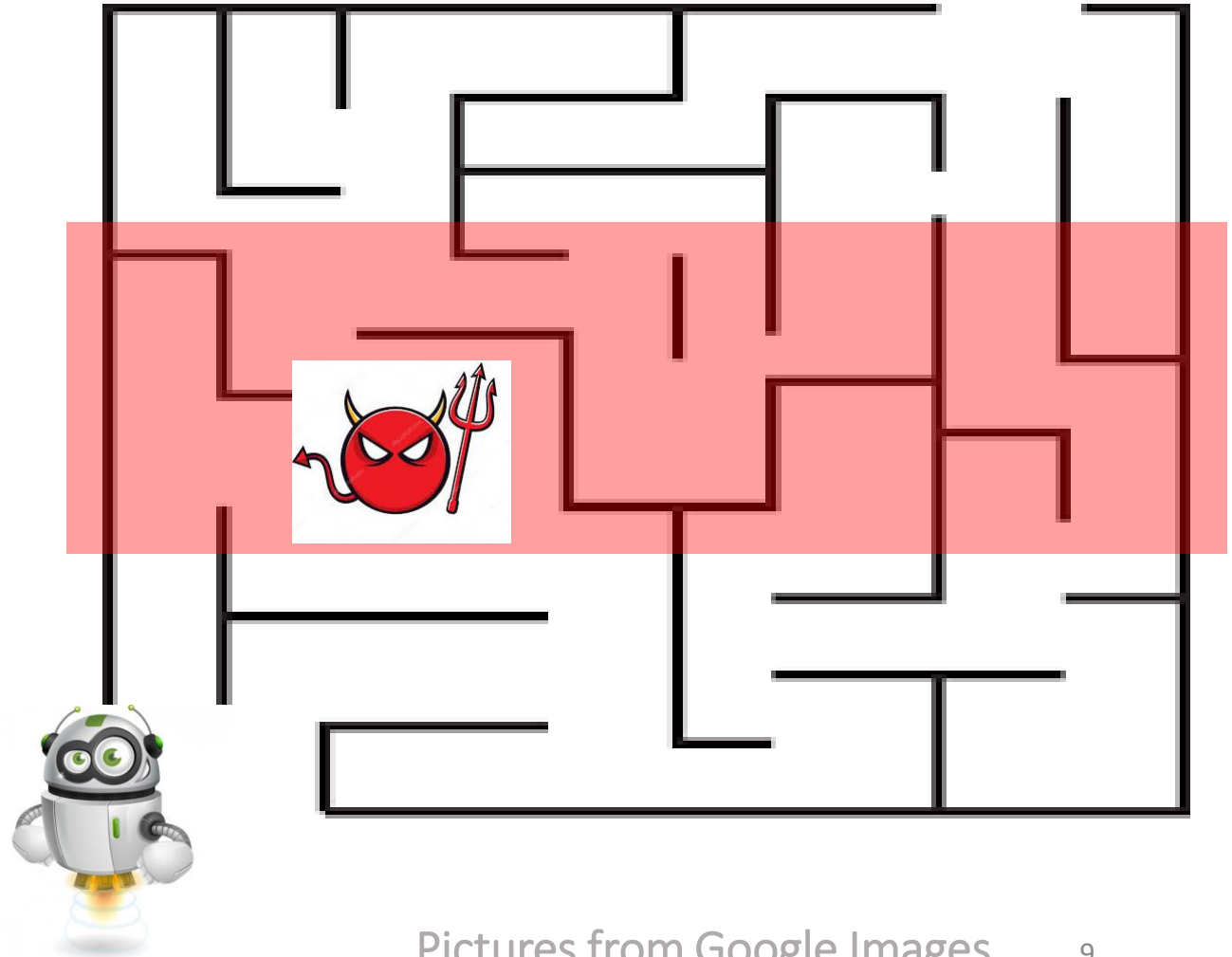


# Let's help the robot plan

Robot in an uncontrollably changing environment

Robot must satisfy a given specification

**Reactive synthesis to the rescue!**



# Synthesis from **rich** specifications

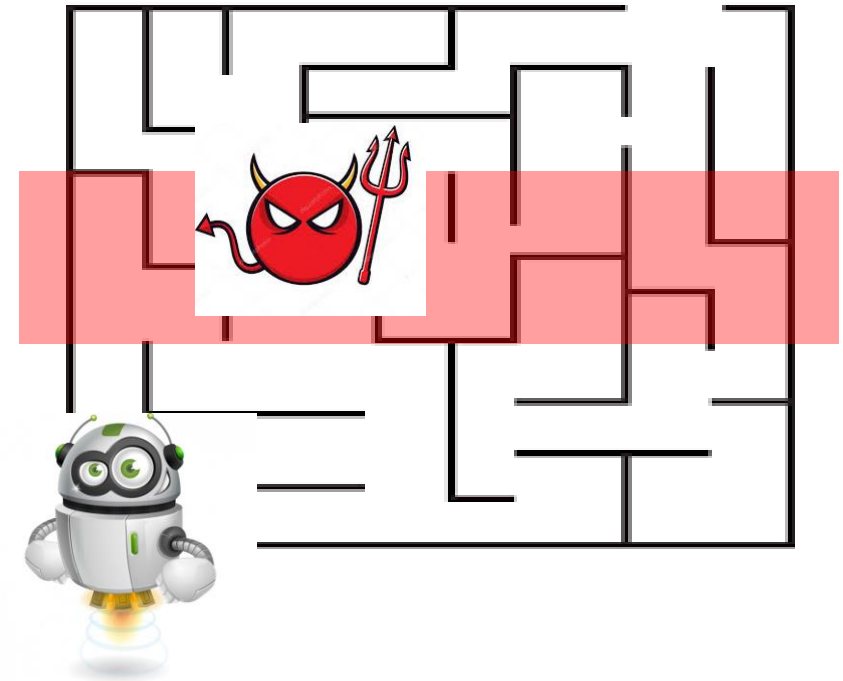
**Specification = Quantitative + Qualitative**

## **Qualitative: Temporal Goals**

Given an **LTL formula**, every execution should satisfy the formula

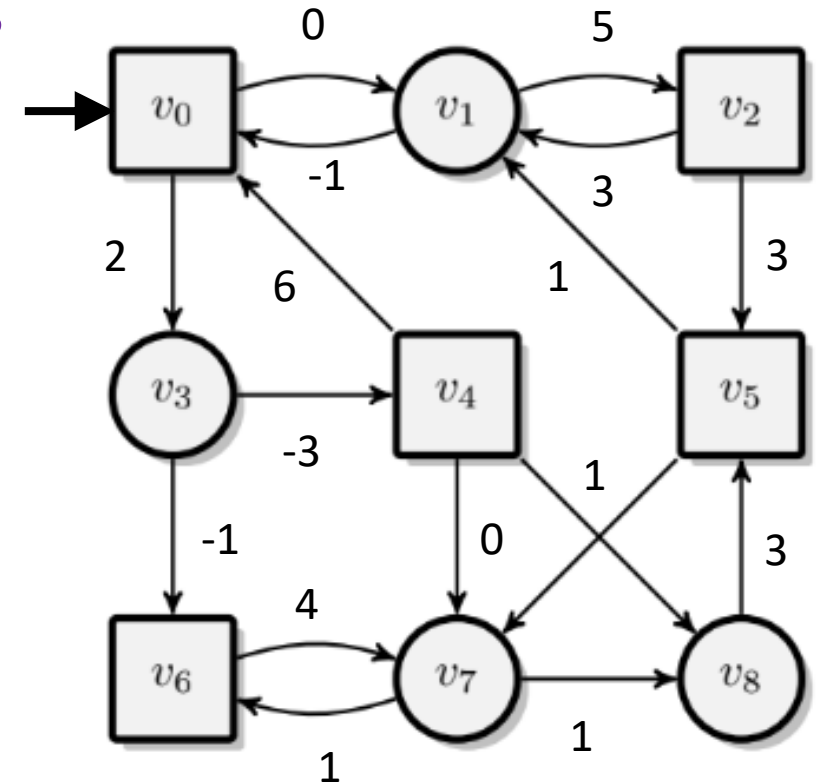
## **Quantitative: Satisficing Goals**

Given a **threshold value**, cost of every execution should exceed the threshold



# Quantitative Game

- Two-player graph game with weights on edges
  - Plays begin in initial state; From each state, its player chooses the next state
- Cost of a play (Discounted-sum):
  - For weight sequence  $A$  and discount factor  $d > 1$ ,
$$DS(A, d) = A[0] + \frac{A[1]}{d} + \frac{A[2]}{d^2} + \dots$$
- Adversarial players
  - System player: Maximizes cost of plays
  - Environment player: Minimizes cost of plays



# Synthesis from Temporal and Satisficing Goals

**Strategy:** Decides the next state based on the history of a play

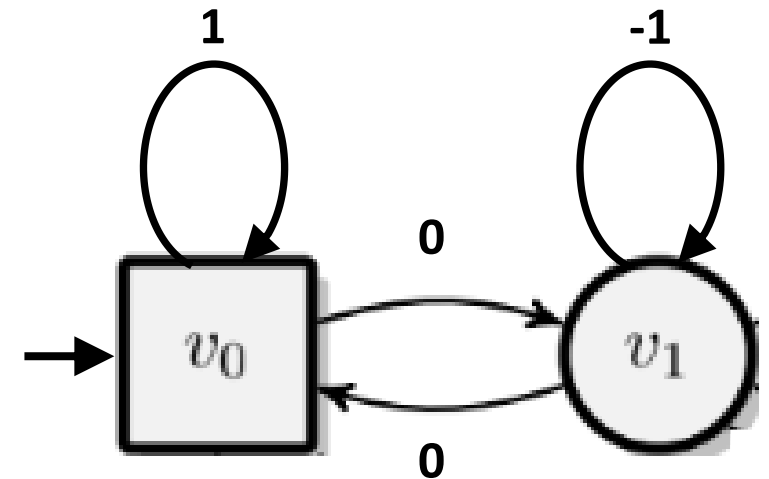
**Problem:** Generate a strategy for the system player that

- (a). satisfies a given LTL formula on all plays
- (b). ensures the cost of all plays exceeds a given threshold value.

## Example

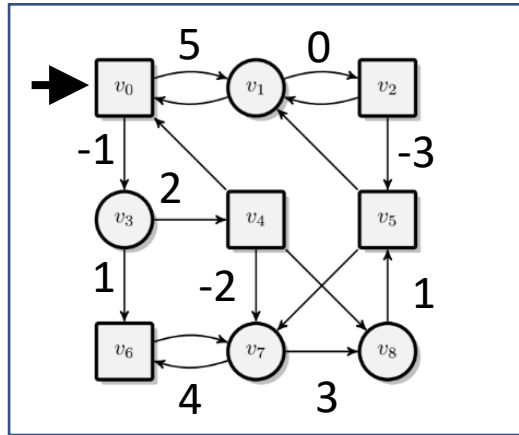
LTL Goal: Visit state  $v_1$

Satisficing Goal: Ensure cost exceeds 0.5



# Existing Solution Approaches

[Chatterjee et. al. 2017; Wen, Ehlers, Topcu, 2015; Kwiatkowska, Parker, Wiltsche; 2017]



Quantitative Game  
 $d > 1$

Temporal  
Goal



Automata-  
based  
methods

Satisficing  
Goal



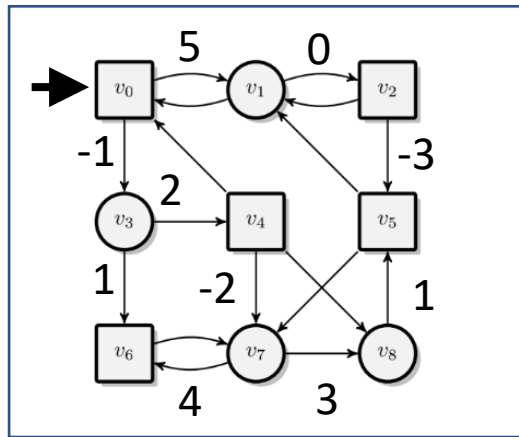
Optimization-  
based  
methods

**No sound algorithm so far!**

Disparate methods do not combine well

# Existing Solution Approaches

[Chatterjee et. al. 2017; Wen, Ehlers, Topcu, 2015; Kwiatkowska, Parker, Wiltsche; 2017]



Quantitative Game  
 $d > 1$

LTL  
Formula



Automata-  
based  
methods

Threshold  
Value



Optimization-  
based  
methods

**No sound algorithm so far!**

Disparate methods do not combine well

# Existing Solution Approaches

[Chatterjee et. al. 2017; Wen, Ehlers, Topcu, 2015; Kwiatkowska, Parker, Wiltsche; 2017]

Automata-

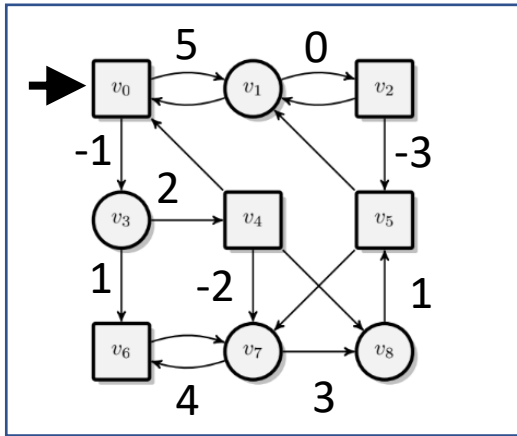
Devise integrated methods that  
combine both steps

$d > 1$

**No sound algorithm so far!**

Disparate methods do not combine well

# Our Approach: Devising Integrated Solutions



Quantitative Game  
 $d > 1$

Temporal  
Goal



Automata-  
based  
methods

Satisficing  
Goal



Automata-based Methods for  
Satisficing Goals?



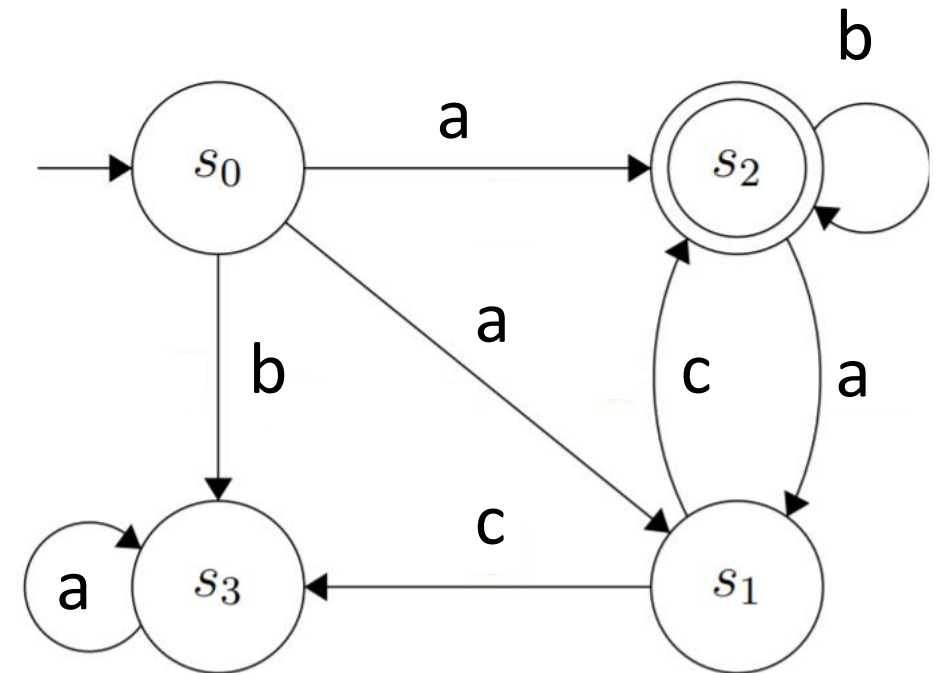
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  - ✓ Existing algorithms fail to offer formal guarantees due to disparate-methods
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  - Towards algorithms with formal guarantees and scalable performance
- Generality of approach in Formal Quantitative Reasoning
  - Beyond reactive synthesis

# Non-deterministic Büchi Automata (NBA)

Automata over infinite-length words

- Finite states and finite alphabet
- Accepting states
- Transitions between states on alphabet
  - Deterministic BA: Exactly one transition from every state on every alphabet
- Run is accepting if it visits accepting states infinitely often
- Word is accepting if it has an accepting run



# Comparison is the fundamental operation

Satisficing Goal: Given, threshold value  $v$

System-player wins

If cost of plays exceeds  $v$

$\equiv DS(A, d) > v$ , where  $A$  is weight-sequence of a play

Given, weight sequences  $A$  and  $B$ , is  $DS(A, d) > DS(B, d)$ ?

How to perform comparison using automata?

# Comparator

[Bansal, Chaudhuri, and Vardi. FoSSaCS 2018; Bansal et. al. CAV 2018; Bansal and Vardi, CAV 2019]

Weight sequences are infinite-length words

- Finite alphabet  $\Sigma = \{-\mu, \dots, \mu\}$  for integer  $\mu > 0$

Given, discount factor  $d > 1$

equality or inequality relation  $R \in \{\leq, <, \geq, >, \neq, =\}$

integer  $\mu > 0$

Let,  $A, B \in \Sigma^\omega$  be two weight sequences

Comparator accepts  $(A, B)$  iff  $DS(A, d) \text{ } R \text{ } DS(B, d)$

# Is Comparator an Automata?

With integer discount factors  $d > 1$

- $DS(A, d) = a_0 + \frac{a_1}{d} + \frac{a_2}{d^2} + \dots = (a_0.a_1a_2\dots)_d = A_d$
- $DS(A, d) \leq DS(B, d)$  iff  $A_d \leq B_d$
- Is there  $C$  s.t.  $DS(C, d) = B_d - A_d \geq 0$ ?

Caveat:  $a_i \geq d$  and -ve

Caveat: Difference of infinite sequences?

# Core Insight

- Consider ( $d = 10$ )

Most significant

Least significant

X

A                    5    13    6

C            +    0    8    6

B

# Core Insight

- Consider ( $d = 10$ )

Most significant ←				→ Least significant
X			1	0
A	5	13	6	
C	+	0	8	6
B			2	

$$i > 0, \quad a_i + c_i + x_i = b_i + d \cdot x_{i-1}$$

# Core Insight

- Consider ( $d = 10$ )

	Most significant	←	Least significant
X	2	1	0
A	5	13	6
C	+ 0	8	6
B		2	2

$$i > 0, \quad a_i + c_i + x_i = b_i + d \cdot x_{i-1}$$



# Core Insight

- Consider ( $d = 10$ )

Most significant		←			Least significant
X		2	1	0	
A		5	13	6	
C	+	0	8	6	
B		7	2	2	

$$i = 0, \quad a_0 + c_0 + x_0 = b_0$$

$$i > 0, \quad a_i + c_i + x_i = b_i + d \cdot x_{i-1}$$

# Core Insight

- Consider ( $d = 10$ )

Most significant

Less significant

X							
A	5	13	6	0	0	0	....
C							
B	7	2	2	0	0	0	....

$$i = 0, \quad a_0 + c_0 + x_0 = b_0$$

$$i > 0, \quad a_i + c_i + x_i = b_i + d \cdot x_{i-1}$$

# Core Insight

- Consider ( $d = 10$ )

Most significant		Less significant					
X	2						
A	5	13	6	0	0	0	....
C	+	0					
B	7	2	2	0	0	0	....

$$i = 0, \quad a_0 + c_0 + x_0 = b_0$$

$$i > 0, \quad a_i + c_i + x_i = b_i + d \cdot x_{i-1}$$

# Core Insight

- Consider ( $d = 10$ )

Most significant		Less significant				
X	2	1				
A	5	13	6	0	0	0 ....
C	+	0	8			
B	7	2	2	0	0	0 ....

$$i = 0, \quad a_0 + c_0 + x_0 = b_0$$

$$i > 0, \quad a_i + c_i + x_i = b_i + d \cdot x_{i-1}$$

# Core Insight

- Consider ( $d = 10$ )

Most significant		Less significant					
X	2	1	0				
A	5	13	6	0	0	0	....
C	+	0	8	6			
B	7	2	2	0	0	0	....

$$i = 0, \quad a_0 + c_0 + x_0 = b_0$$

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# Core Insight

- Consider ( $d = 10$ )

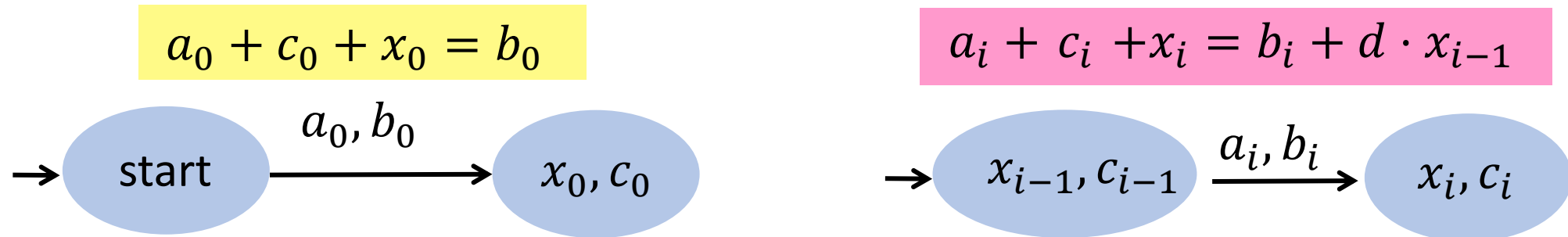
		Most significant		Less significant				
X		2	1 0	0	0	0	....	
A		5	13	6	0	0	0	....
C	+	0	8	6	0	0	0	....
B		7	2	2	0	0	0	....

$$i = 0, \quad a_0 + c_0 + x_0 = b_0$$

$$i > 0, \quad a_i + c_i + x_i = b_i + d \cdot x_{i-1}$$

$$\rightarrow DS(A, d) + DS(C, d) = DS(B, d)$$

# Comparator: Construction



$x_i, c_i$  are bounded integers, so automaton has finitely many states

## Theorem

DS comparator for (integer)  $d, \mu, R$  is an NBA with  $O(\mu^2)$  states

# Is Comparator an Automata

- With integer discount factors  $d > 1$ 
  - $DS(A, d) = a_0 + \frac{a_1}{d} + \frac{a_2}{d^2} + \dots = (a_0.a_1a_2\dots)_d = A_d$
  - $DS(A, d) \leq DS(B, d)$  iff  $A_d \leq B_d$  Caveat:  $a_i \geq d$  and -ve
  - Is there  $C$  s.t.  $DS(C, d) = B_d - A_d \geq 0$ ? Caveat: Difference of infinite sequences?
  - Non-determinism for arithmetic from most to lesser significant digits
- With non-integer discount factors, comparator is not an automata

**Theorem:** Comparator is an NBA iff the discount-factor is an integer



# Satisficing via Comparators

Satisficing Goal: Given, threshold value  $v$

System-player wins

If cost of plays exceeds  $v$

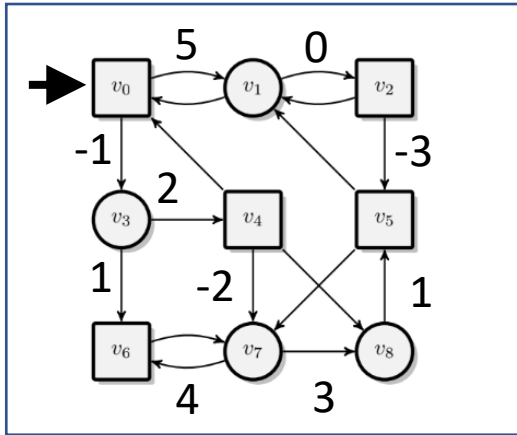
$\equiv DS(A, d) > v$ , where  $A$  is weight-sequence of a play

When the discount factor is an integer

$\equiv (A, V)$  is accepted by a comparator for  $>$  where  $DS(V, d) = v$

$\equiv$  **Comparator for  $>$**  captures winning condition

# Our Approach: Devising Integrated Solutions



Quantitative Game  
Integer  $d > 1$

Temporal  
Goal



Automata-  
based  
methods

Satisficing  
Goal

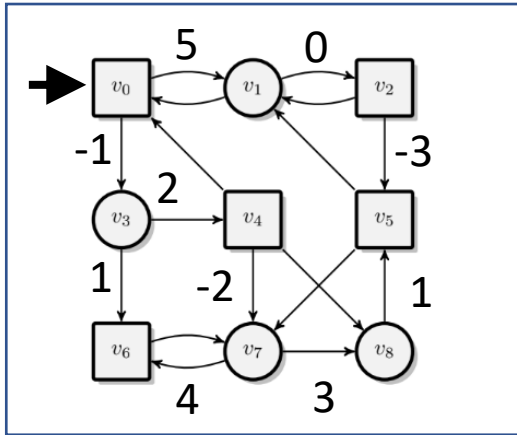


Comparator-  
based  
methods

**Algorithm is sound!**

Automata-based methods combine well!

# Our Approach: Devising Integrated Solutions



Quantitative Game  
**Integer  $d > 1$**

Temporal  
Goal



Automata-  
based  
methods

Satisficing  
Goal



Comparator-  
based  
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**Algorithm is sound!**

Automata-based methods combine well!

# Approximate Comparator

[Bansal, et. al. 2021]

Given, non-integer discount factor  $d > 1$ ,  
approximation factor  $0 < \varepsilon < 1$

Approximate comparator accepts  $(A, B)$ , then

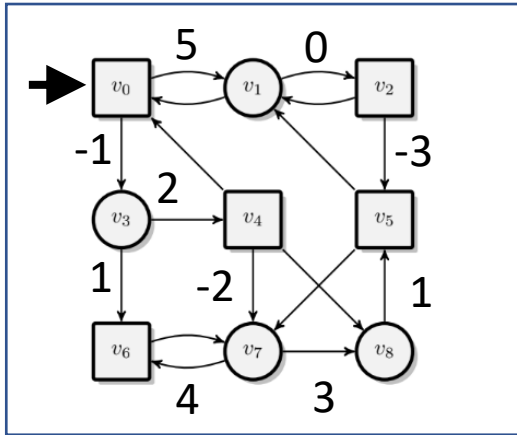
$$DS(A, d) > DS(B, d)$$

Approximate comparator rejects  $(A, B)$ , then

$$DS(A, d) \leq DS(B, d) + d \cdot \varepsilon$$

**Theorem:** Approximate comparator is an NBA (under some assumptions)

# Our Approach: Devising Integrated Solutions



Quantitative Game  
 $d > 1$

Temporal  
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Automata-  
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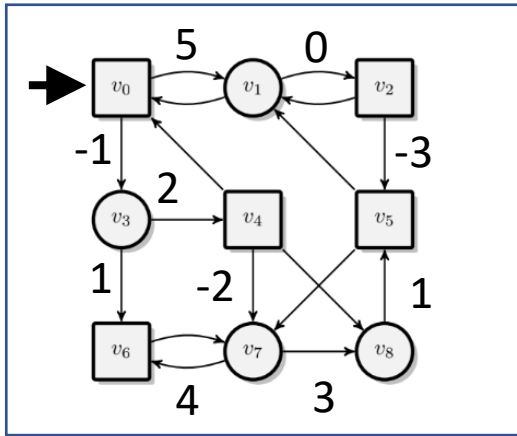
Satisficing  
Goal



Approx.  
comparator- or  
comparator-based  
methods

**First theoretically sound algorithm for  
Synthesis from qualitative and qualitative goals**

# Our Approach: Theoretical sound but ....



Quantitative Game  
 $d > 1$

**... not scalable**

Temporal  
Goal



Automata-  
based  
methods

Satisficing  
Goal



Approx.  
comparator- or  
comparator-based  
methods

All are  
NBAs



Deterministic  
NBAs

**Exponential blow-up!**

# Deterministic Comparators

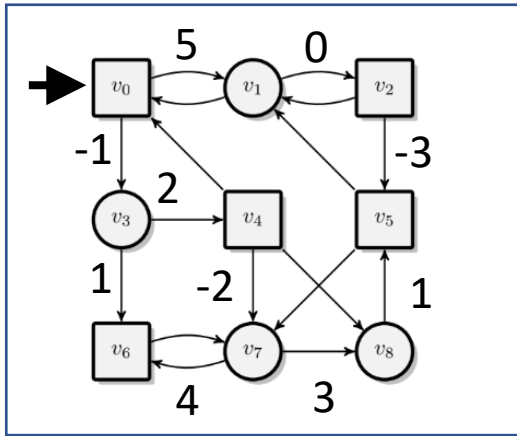
[Bansal and Vardi. CAV 2019]

**Theorem:** Comparators and Approximate Comparators are **safety/co-safety automata**

**Safety and co-safety automata** are deterministic NBAs that focus on finite prefixes

- Safety rejects words based on a finite prefix
- Co-safety accepts words based on a finite prefix
- $DS(A, d) = a_0 + \frac{a_1}{d} + \frac{a_2}{d^2} + \dots = DS(A[0 \dots i], d) + \frac{1}{d^i} \cdot DS(A[i \dots], d)$
- As  $i$  increases, **Tail**  $\rightarrow 0$ . Eventually, only the finite prefix  $A[0 \dots i]$  matters!

# Our Approach: Theoretically sound and ....



Quantitative Game  
 $d > 1$

Temporal  
Goal



Automata-  
based  
methods

Satisficing  
Goal



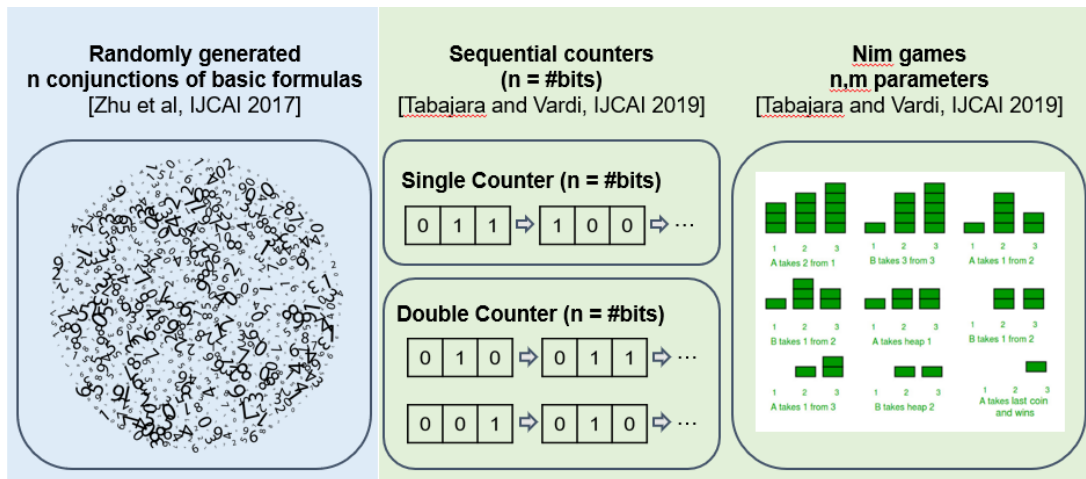
**Deterministic Approx.**  
comparator- or  
comparator-based  
methods

... scalable in practice



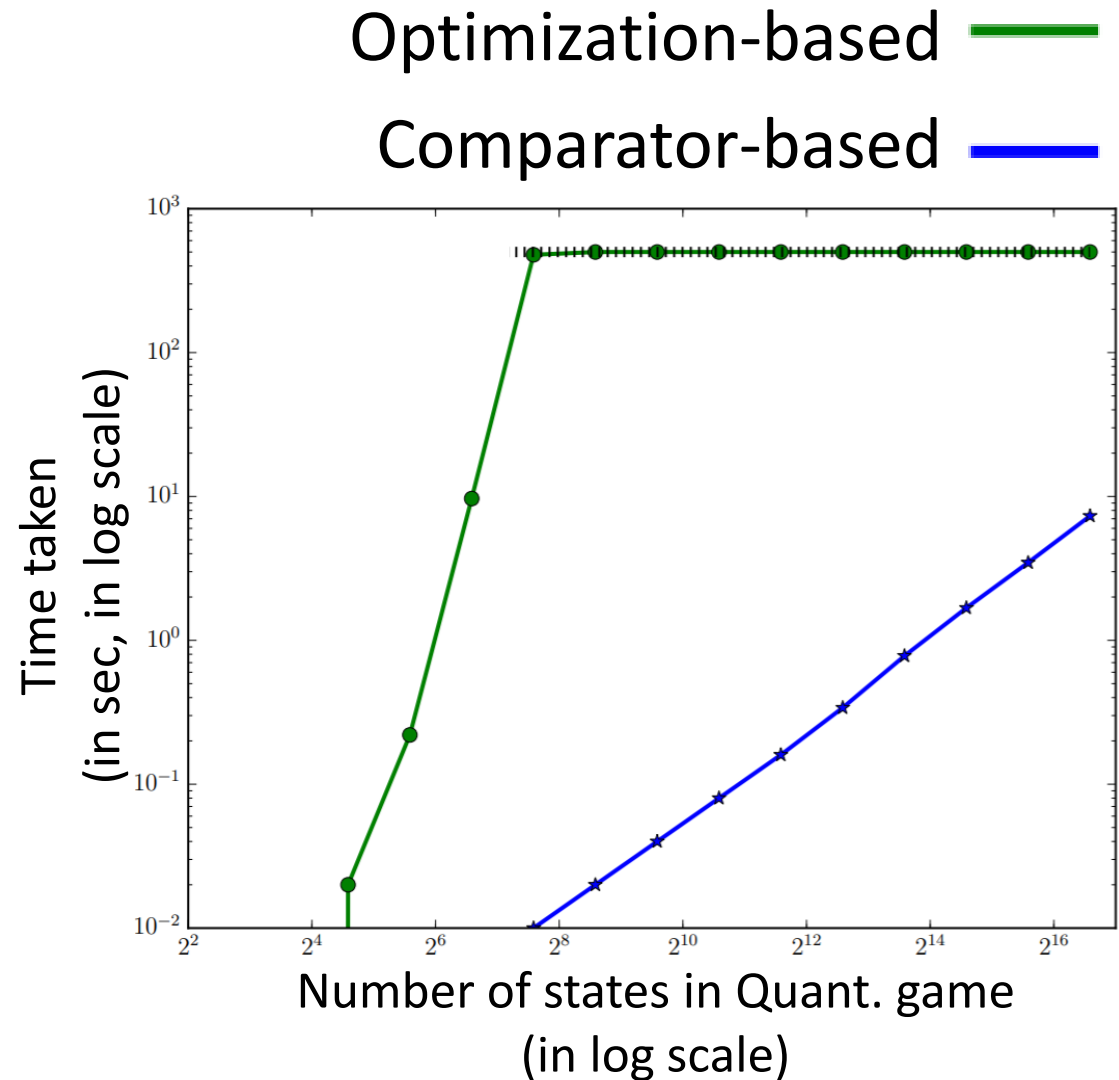
# Synthesis from Satisficing Goals only: Optimization vs Comparators

300 benchmarks created from  
suite of LTLf benchmarks

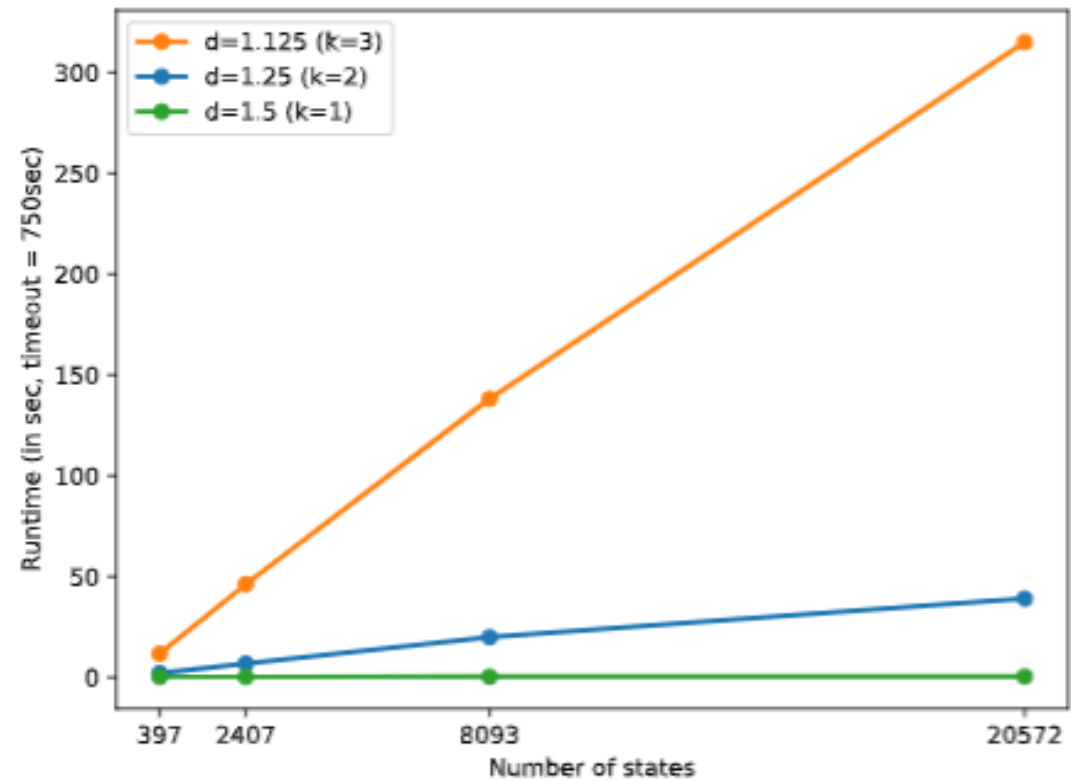
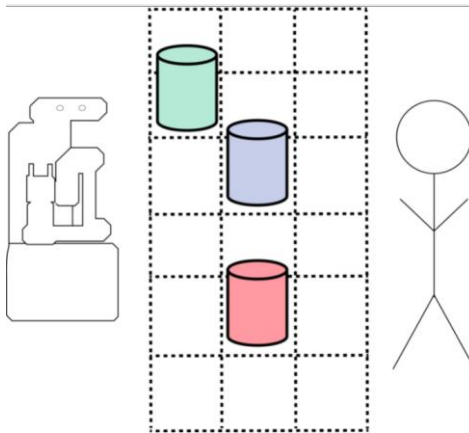
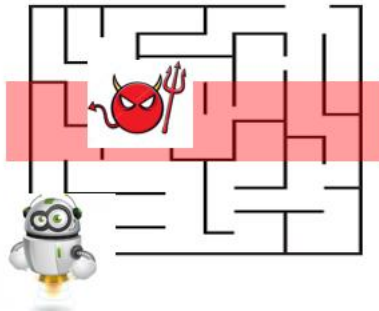


Optimization:  $\sim 140/300$

Comparator:  $\sim 285/300$



# Synthesis from Temporal and Satisficing Goals



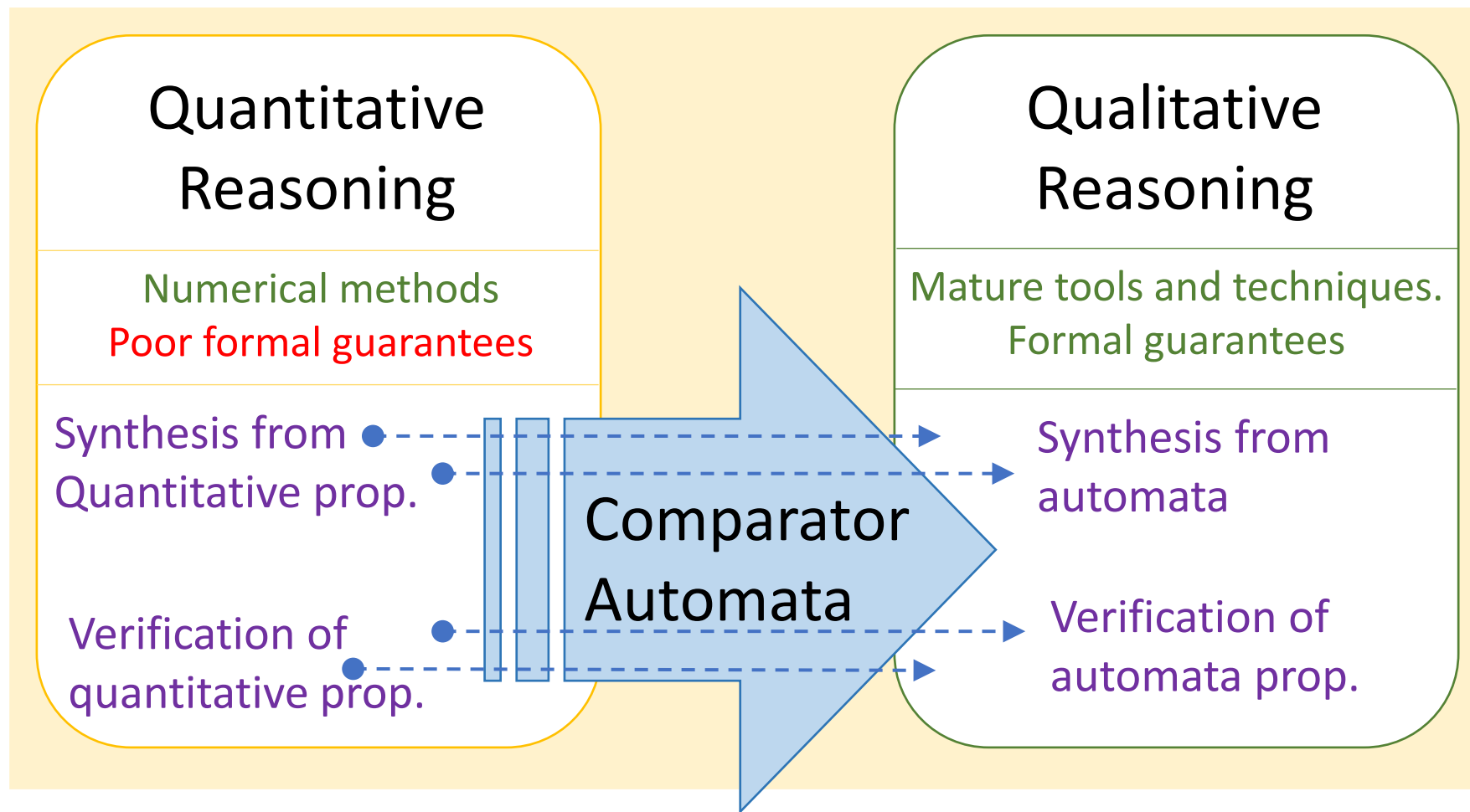
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  - ✓ Existing algorithms fail to offer formal guarantees due to disparate-methods
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  - ✓ Comparator-based algorithms
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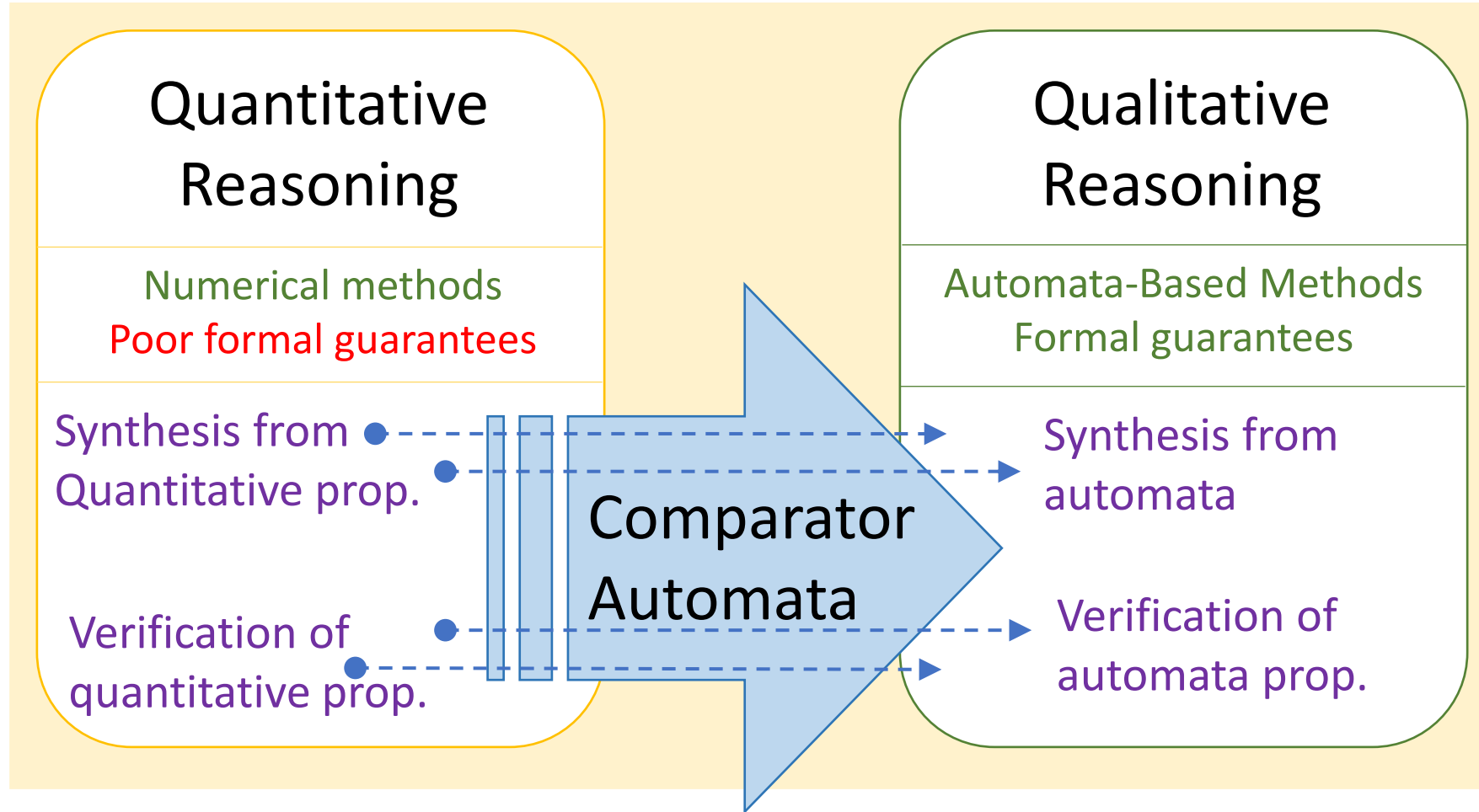
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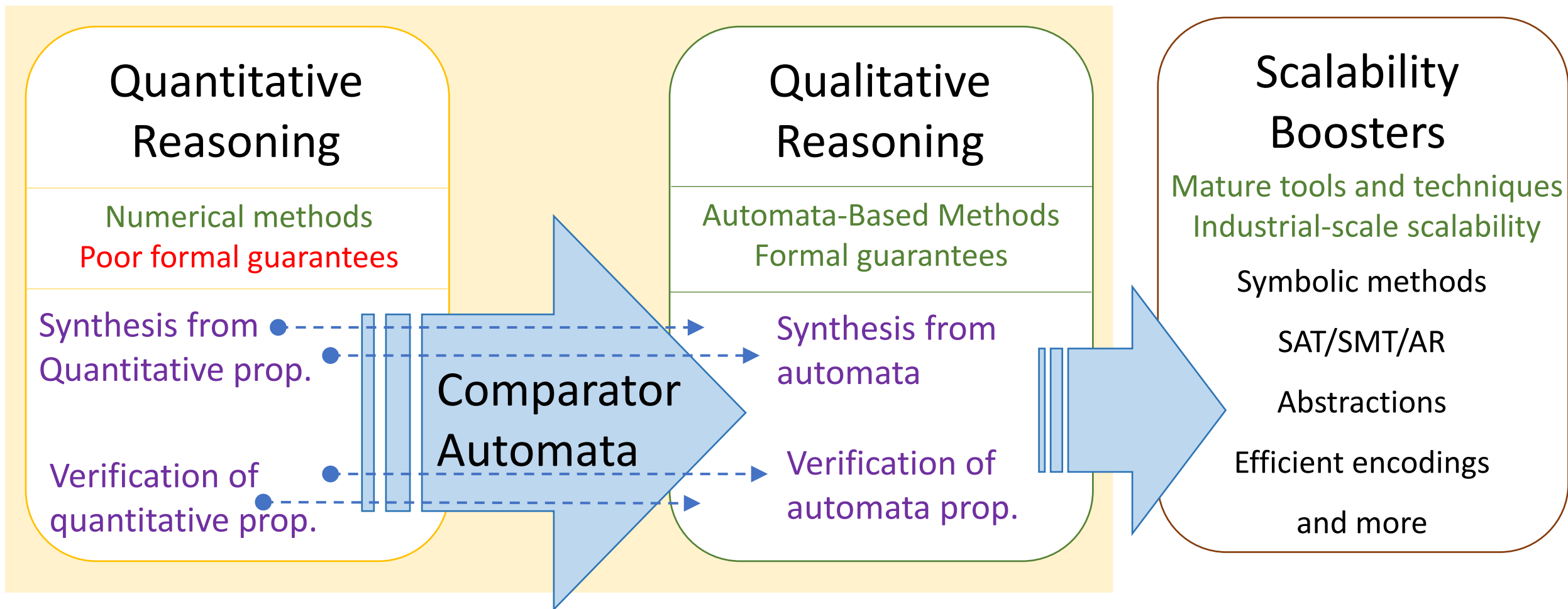
# Automata-Based Quantitative Reasoning



# Automata-Based Quantitative Reasoning



# Automata-Based to Constraint Solving



Real-life applications in planning (robotics), formal guarantees on reinforcement learning

# Towards designing intelligent systems via reactive synthesis

- Reactive synthesis from rich specifications
  - This talk: Quantitative + Qualitative specifications
- Automata-theoretic quantitative reasoning
    - Formal guarantees and scalability
    - Combine well with qualitative reasoning
- Beyond synthesis
    - Principled study of comparators and their capabilities
    - Quantitative reasoning -> Constraint solving



# Many thanks to my collaborators



Swarat Chaudhuri  
Rice U



Krishnendu Chatterjee  
IST Austria



Moshe Y. Vardi  
Rice U



Lydia Kavraki  
Rice U



Andrew Wells  
Tesla

# Towards designing intelligent systems via reactive synthesis

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