

On Satisficing in Quantitative Games

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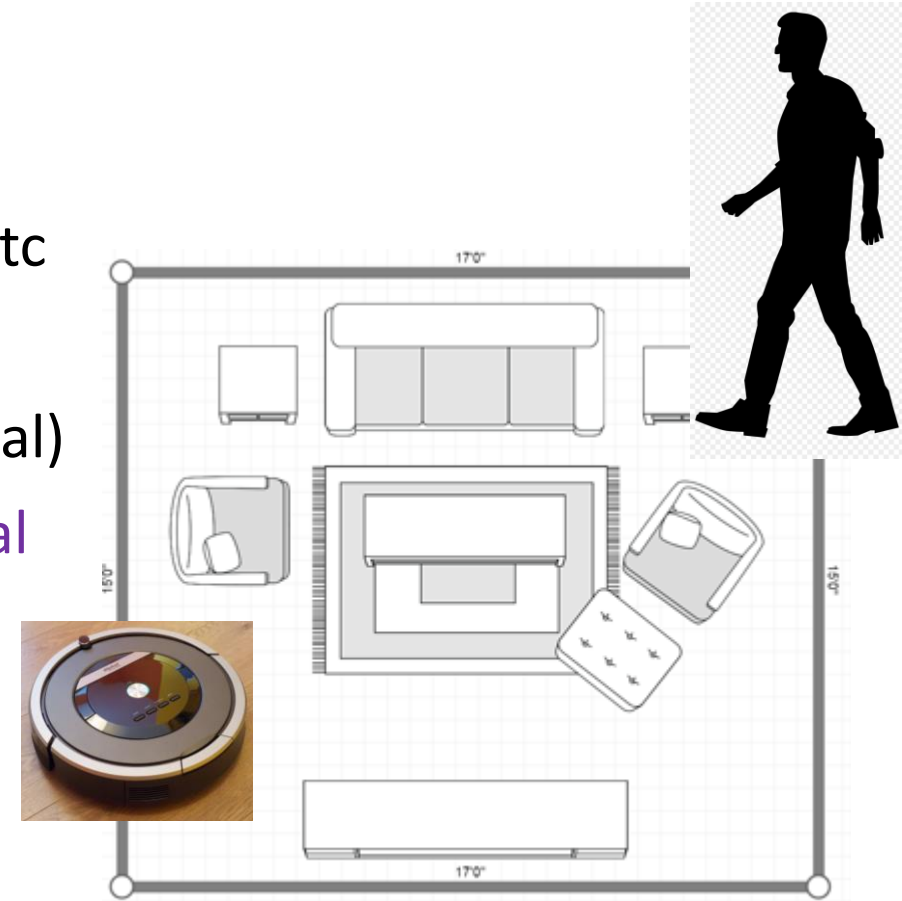
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Synthesis from Quantitative Constraints

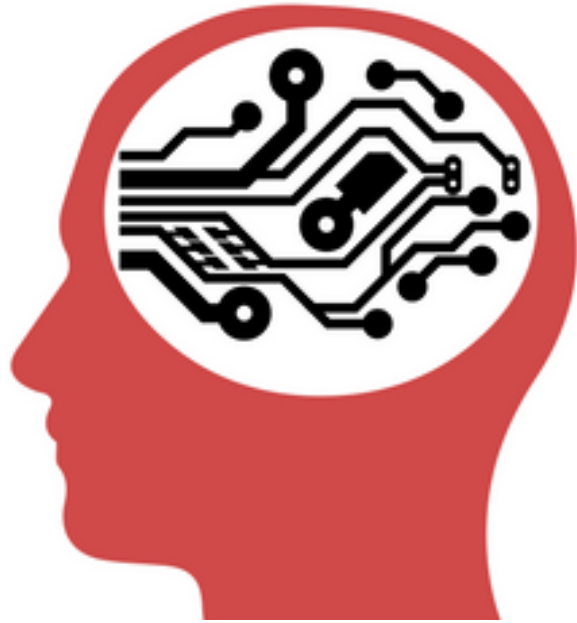
Quantitative constraints

- Resource consumption, quality measures, rewards, etc
- Controller with **minimal** battery consumption (Optimal)
- Controller ensures battery **never goes below a critical threshold** (good-enough)

Satisficing: To search for “good enough” solutions
- Herb A. Simons



Benefits of satisficing



Computationally Easier

Quantitative
+
Temporal

Amenable to
combination of
constraints

In this paper:

Formulate the Satisficing Problem in Quantitative Games

Approach 1: Via Optimization

- Inherits deficiencies of optimization ☹️

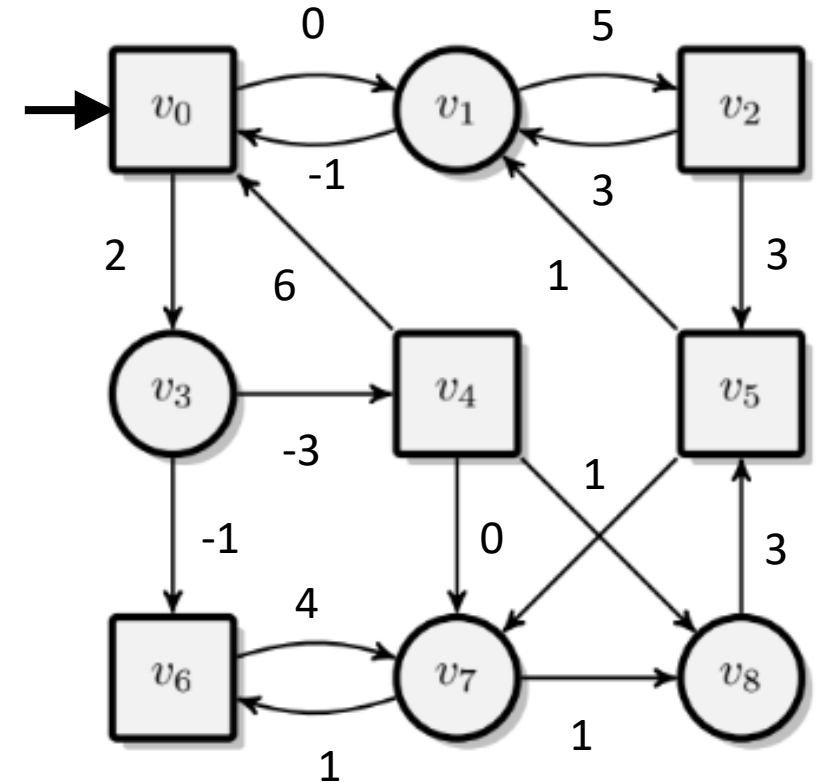
Approach 2: Via Automata-based methods

- Demonstrates theoretical and empirical benefits

Quantitative Game

- States partitioned into two types
 - Each player owns one type of states
- Play
 - Begins 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 $d > 1$,

$$DS(A, d) = A[0] + \frac{A[1]}{d} + \frac{A[2]}{d^2} + \dots$$



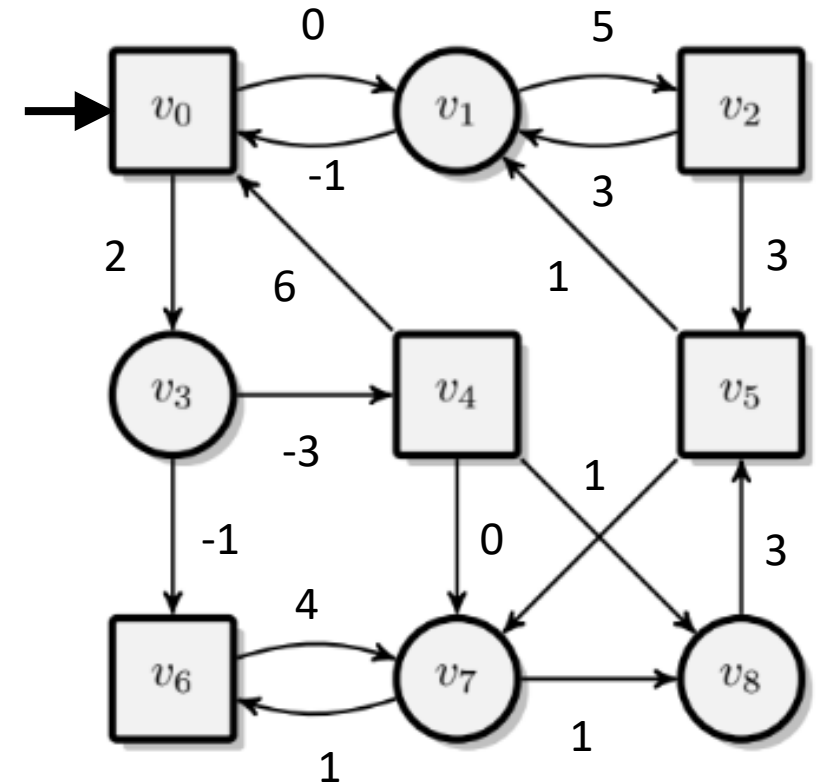
Quantitative Game (cont...)

- Adversarial players

- Max-player: To maximize cost of plays
- Min-player: To minimize cost of plays

- Strategy for a player

- Determines which state to go next based on the history of a play?



Satisficing Problem

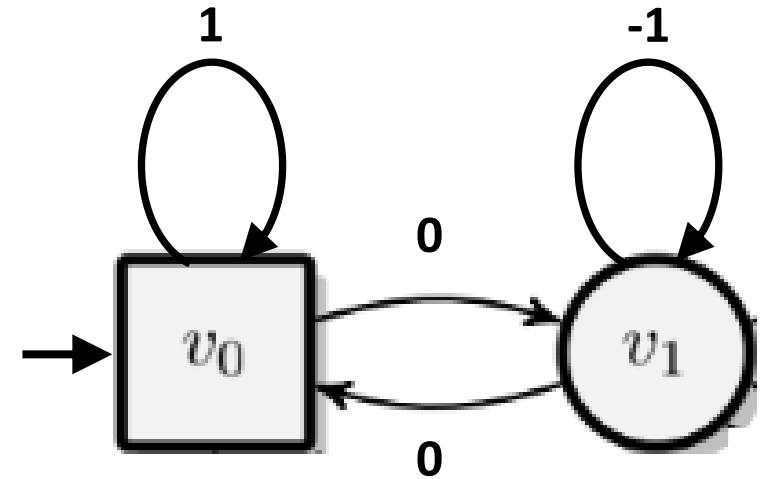
- Strategies that obey a threshold
- Threshold could be a physical constraint

Satisficing Problem

Given a threshold value v ,
Does the Max-Player have a strategy
such that cost of all plays exceeds v ?

Example: Threshold value 0.5

“Max-player wins”



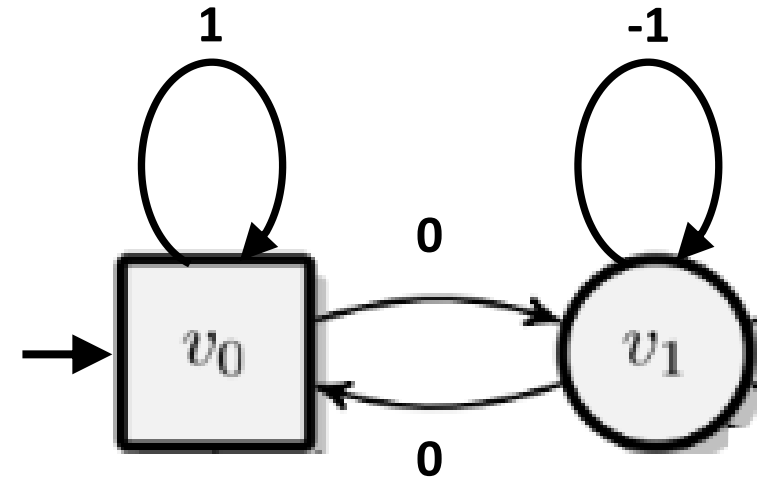
Satisficing via Optimization

Optimization

- To compute the **optimal cost**
 - Optimal cost is when both players play optimally
 - Pseudo-polynomial time [Shapely 1950]

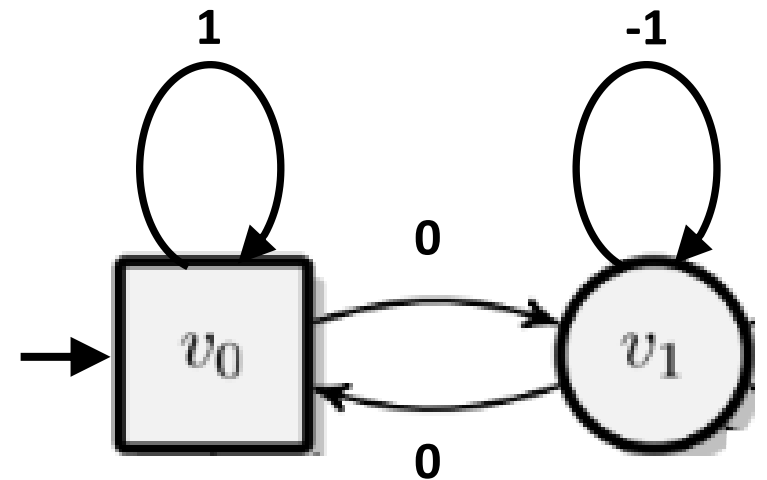
Satisficing via Optimization

- Max-player wins iff $\text{Threshold} \leq \text{Optimal Cost}$



Satisficing via Optimization: Drawbacks

- Same worst-case complexity as optimization
 - No improvement in complexity
- Unsound for Satisficing + Temporal Goals
 - Satisficing threshold: 0.5
 - Temporal goal: Visit v_1



So far on Satisficing

- Satisficing via Optimization 😞
 - No complexity improvement over optimization
 - Unsound with temporally extended goals
- Quantitative + Qualitative goals is challenging [Kwiatkowska, FSE 2006]
 - Inherent incompatibility of techniques
 - Quantitative uses numerical methods
 - Qualitative uses automata-based methods
- Satisficing via an automata-based method?
 - Eliminate the incompatibility!

Comparison is the fundamental operation

Max-player wins

If cost of plays exceeds v

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

How to perform comparison?

Comparator automata

[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 $\mu > 0$

Comparator automata (comparator) accepts $A \in \Sigma^\omega$ iff $DS(A, d) R 0$

Comparator is ω -regular iff the discount factor is an integer

For integer discount factors,

Comparator is a deterministic safety/co-safety automata with $O(\mu)$ states

Satisficing via Comparators

Max-player wins

If cost of plays exceeds 0

$\equiv DS(A, d) > 0$, where A is weight-sequence of a play

When the discount factor is an integer

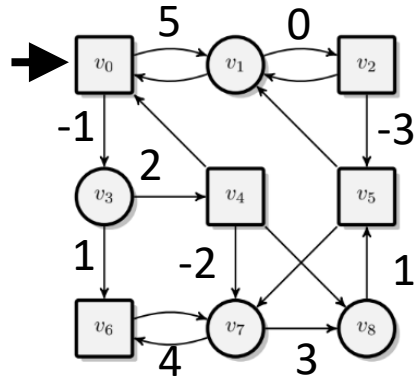
$\equiv A$ is accepted by a comparator for $>$

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

\equiv **Safety/co-safety automata** captures winning condition

Satisficing via Comparators: Reduction

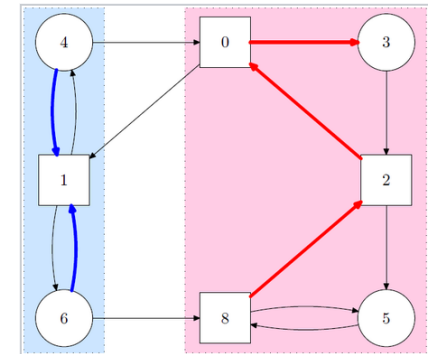
Threshold value 0



Integer $d > 1$
Maximum of absolute
value of weights μ



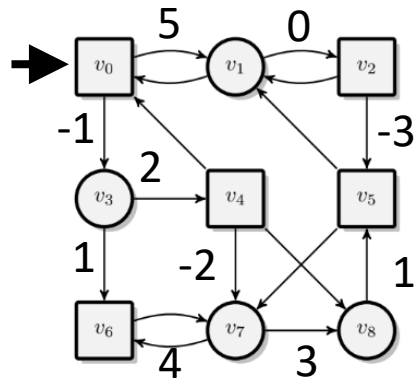
Deterministic
safety/co-safety
automata
for
Comparator for d, μ, R



**Safety
or
reachability
game**

Satisficing via Comparators: Reduction

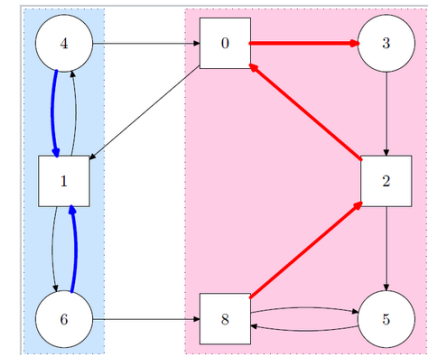
Threshold value 0



Integer $d > 1$
Maximum of absolute
value of weights μ



Deterministic
safety/co-safety
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for
Comparator for d, μ, R



**Safety
or
reachability
game**

Comparator automata with arbitrary thresholds

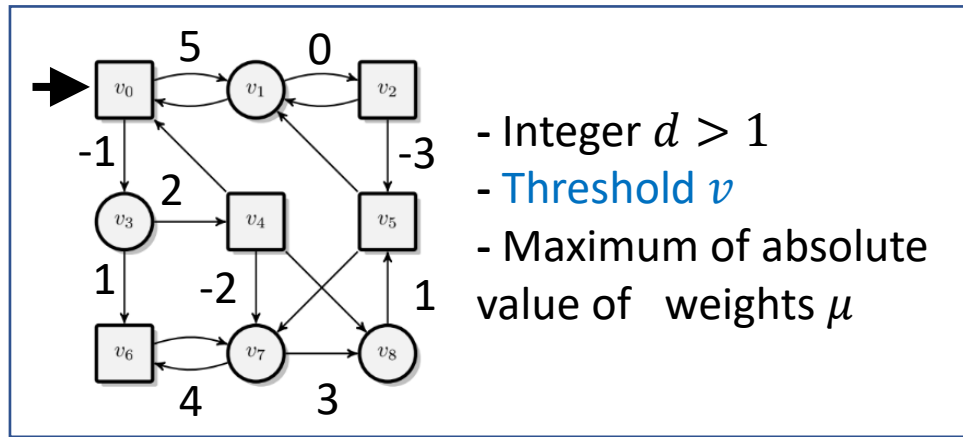
- Comparator automata (comparator) for d, R, μ, v
 - Rational value v
 - Accepts weight sequence $A \in \Sigma^\omega$ iff $DS(A, d) R v$

Comparator is ω -regular iff the discount factor is an integer

Comparator for an integer discount factor is a deterministic safety/co-safety automata with $O(\mu \cdot n)$ states where n is a parameter in v

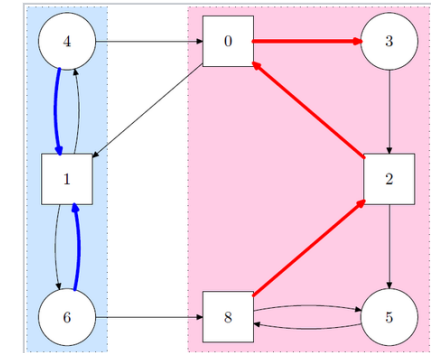
- Rational number has a lasso-word representation
- Product of comparator with lasso-word for threshold doesn't work
 - Transition on $a \times$ Transition on $b \rightarrow$ Transition on $a+b$?
 - Distorts the alphabet from $\{-\mu, \dots, \mu\}$ to $\{-\mu + k, \dots, \mu + k\}$

Satisficing via Comparators: Algorithm



Synchronized
product with

**Deterministic safety
or co-safety
automata** for
comparator for
 d, μ, R, v



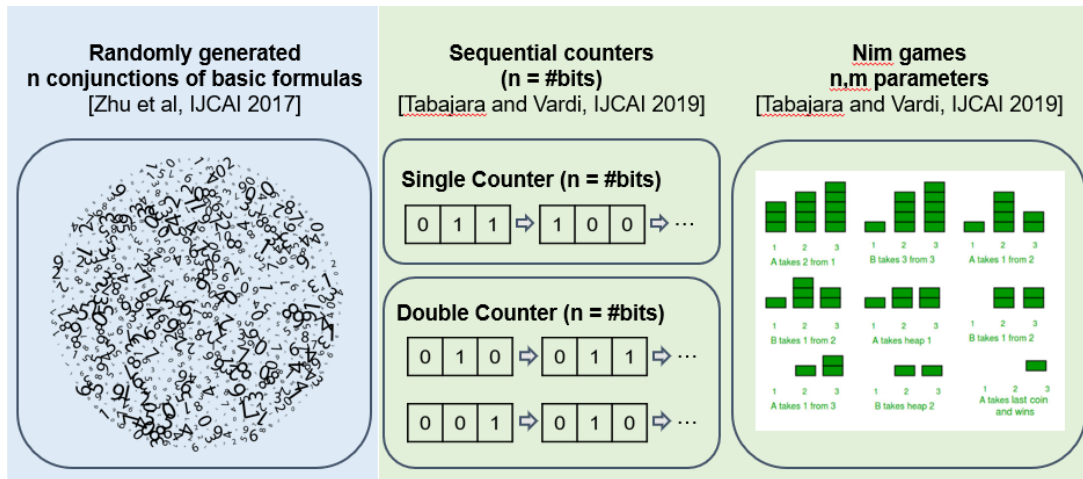
Safety or reachability game

For integer discount factors $d > 1$,

1. Satisficing reduces to solving a safety/reachability game
2. Satisficing takes $O((|V| + |E|) \cdot \mu \cdot n)$
where threshold v has an n -bit representation

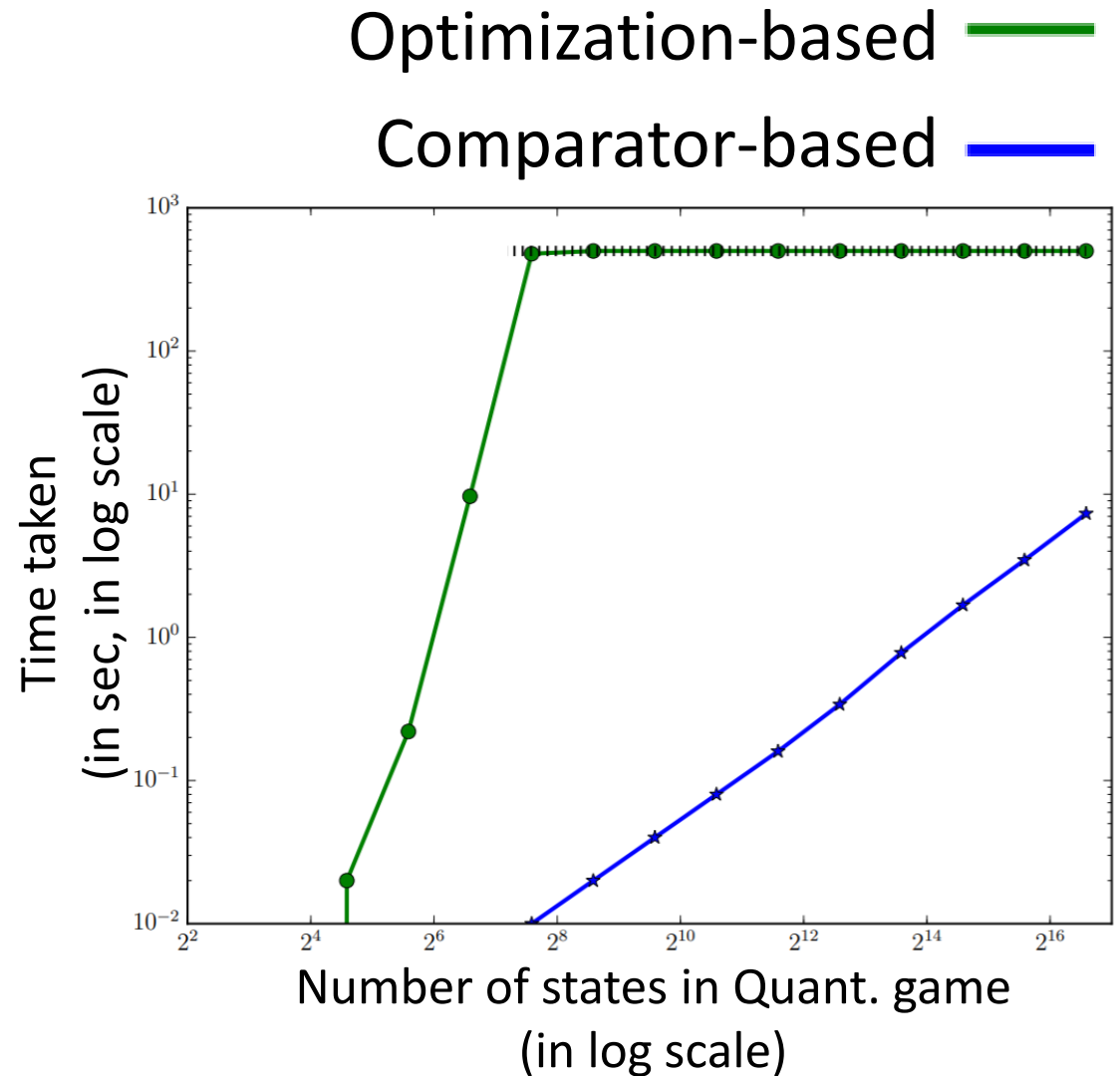
Optimization vs Comparators - I

300 benchmarks created from
suite of LTLf benchmarks

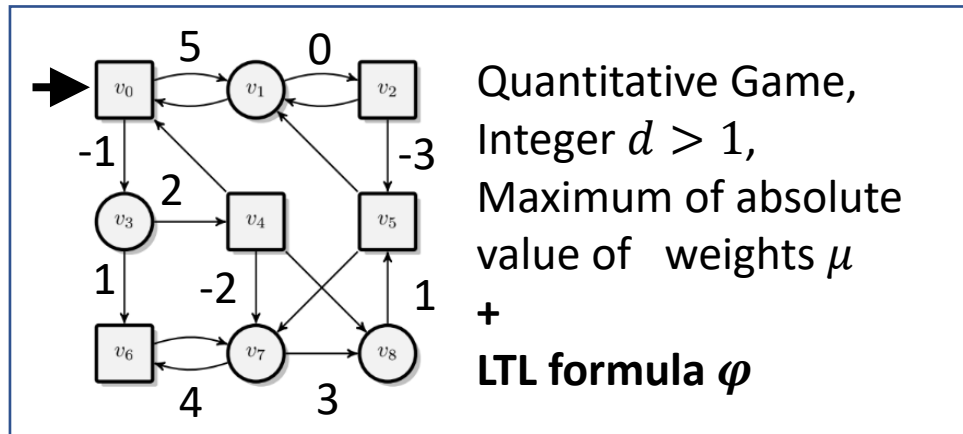


Optimization: $\sim 140/300$

Comparator: $\sim 285/300$



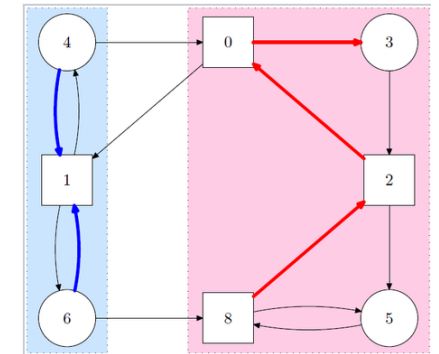
Optimization vs Comparators - II



Synchronized
product with

**Deterministic safety
or co-safety
automata** for DS
Comparator for
 $d, \mu, >, v$

**Deterministic Parity
Automata** for LTL
formula φ



Parity Game

Winning in Parity Game iff
Exists satisficing strategy that satisfies the LTL formula

In a nutshell

On Satisficing in Quantitative Games

- Satisficing is an effective alternative to optimization
- Comparator-based approach is scalable, efficient, has broad applicability
- Fractional discount factors [Under Submission]
 - Comparators for (additive) approximations of DS
 - Applications in planning under adversarial domains
- Algorithmic improvements: Leverage progress in qualitative reasoning
- Satisficing in other domains: MDPs, w/o partial information

Back-up slides

Optimization vs Comparators - III

Optimization-based

- Value Iteration
- Early-termination Heuristic

Comparator-based exhibits robust performance

Optimization-based (Heuristic) —

Comparator-based —

