

The CEBug project

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ROCKS project workshop

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- 1 Stochastic counterexamples
- 2 First CEBug example:
Path-based counterexamples via bounded model checking
- 3 Second CEBug example:
Counterexamples as minimal critical subsystems

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Counterexample generation for stochastic systems using bounded model checking

DTMC/MDP/MRM model checking

→ property is violated

→ counterexample

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DTMC model checking \rightarrow property is violated \rightarrow counterexample

A discrete-time Markov chain (DTMC) is a tuple

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A **discrete-time Markov chain (DTMC)** is a tuple

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↑
finite state space

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↑
initial state $s_{init} \in S$

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transition probability matrix $P : S \times S \rightarrow [0, 1]$

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state labeling function $L : S \rightarrow 2^{AP}$

- linear equation system
- SCC-based model checking
- simulation
- ...

DTMC model checking → property is violated → counterexample

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Counterexamples for **DTMCs** and probabilistic properties:

- **not provided** by model checking
- consist of a large (maybe infinite) **set of paths**

Some state-of-the-art **methods**

- **shortest path** (*Damman, Han, and Katoen 2008*) (*Günther, Schuster & Siegle, 2010*)
- **SCC abstraction** (*Andrés, D'Argenio and van Rossum, 2008*)

Counterexamples are **represented** as

- **sets of paths, path trees,**
- **regular expressions**

In CEBug we develop further counterexample generation and representation techniques.

In this talk: two areas based on

- SAT- and SMT-based **bounded model checking**
- SMT- and MILP-based computation of **minimal critical subsystems**

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- reduce the PCTL model checking problem to a **probabilistic reachability** problem
- finite **paths** $\sigma_0, \dots, \sigma_k$ of length k that lead to a target state can be encoded as solutions to the **propositional logic** formula

$$\text{Init}(\sigma_0) \wedge \left(\bigwedge_{i=0}^{k-1} \text{Trans}(\sigma_i, \sigma_{i+1}) \right) \wedge \text{Target}(\sigma_k)$$

- use **SAT-solving** to find solutions with increasing k until the probability mass is sufficient
- important: **loop detection**
- important: **encoding of (MT)BDDs**

SSBMS: SMT-based stochastic bounded model checking

Problem: we find paths with fewer transitions first, instead of paths with higher probabilities.

- transform probabilities on **logarithmic** scale
- consider **transition probabilities** in bounded model checking:

$$\text{Init}(\sigma_0) \wedge \left(\bigwedge_{i=0}^{k-1} \text{Trans}(\sigma_i, \sigma_{i+1}, \hat{p}_i) \right) \wedge \text{Target}(\sigma_k) \wedge \sum_{i=0}^{k-1} \hat{p}_i \geq \log p_t$$

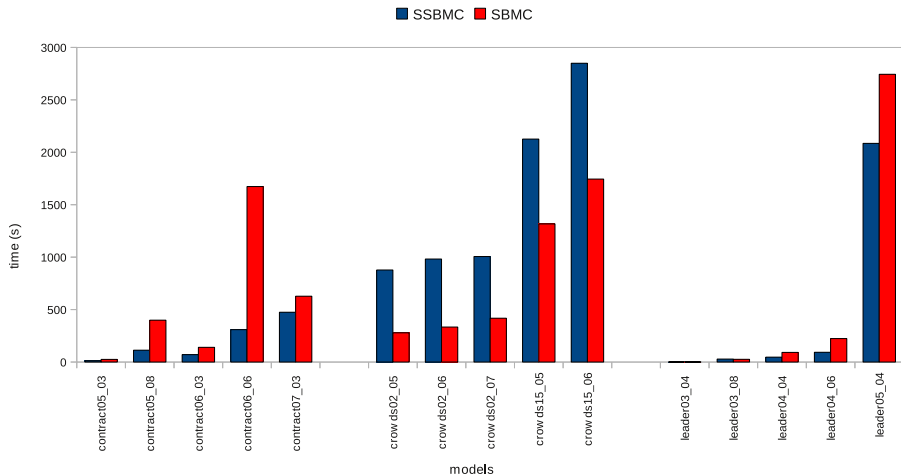
- use **SMT-solving** to find solutions with increasing k / decreasing p_t until the probability mass is sufficient

Experimental Results: SBMC (SAT) vs. SSBMS (SMT)

- **Benchmarks:** Contract signing protocol, Crowds protocol, Leader election protocol, Self-stabilizing minimal spanning tree algorithm
- **Solver:** Minisat, Yices

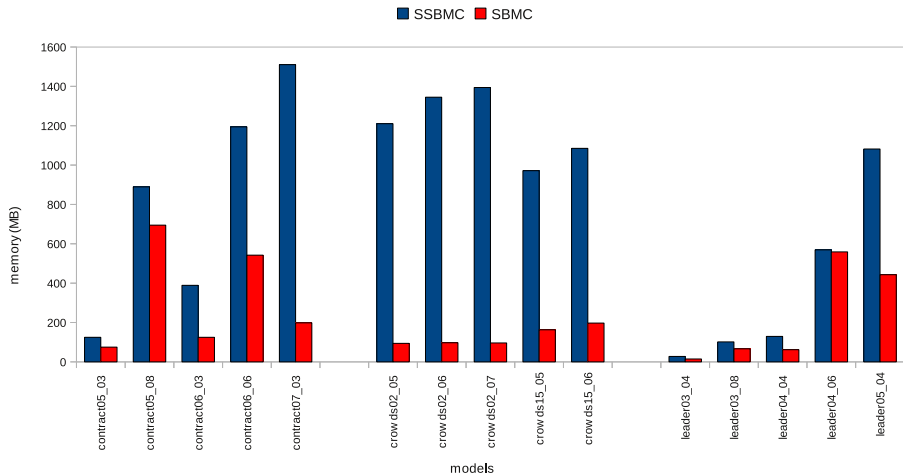
SMT vs. SAT (1)

Computation Time for Contract, Crowds & Leader



SMT vs. SAT (2)

Memory Consumption for Contract, Crowds & Leader



Minimal Spanning Tree:

Name	ρ	k_{\max}	SSBMC			SBMC		
			#paths	time	mem.	#paths	time	mem.
mst15	0.049	15	4531	98.58	148.82	> 600000	- TO -	
mst16	0.047	16	4648	107.27	158.25	> 600000	- MO -	
mst18	0.036	18	4073	109.26	164.24	> 600000	- MO -	
mst20	0.034	20	452	19.57	58.21	> 500000	- TO -	

SAT-search for more probable paths first

Problem: longer but more probable paths are found only after we have searched for shorter paths.

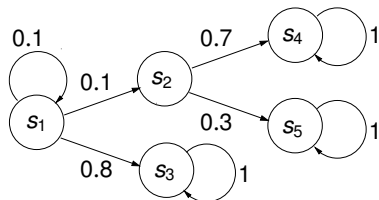
For SAT (the SMT-adaptation is analogous):

- consider τ -transitions in bounded model checking:

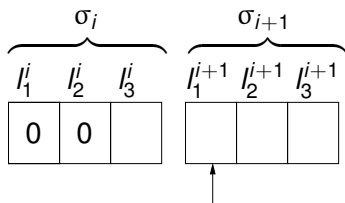
$$Init(\sigma_0) \wedge \left(\bigwedge_{i=0}^{k-1} Trans^{\tau}(\sigma_i, \sigma_{i+1}) \right) \wedge Target(\sigma_k)$$

- modify the SAT-solving to find more probable paths first by an adapted **polarity** heuristics

Searching for more probable paths



	l_1	l_2	l_3
s_1	0	0	0
s_2	0	0	1
s_3	0	1	0
s_4	0	1	1
s_5	1	0	0



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Minimal critical subsystems

- A **subsystem** is a part of a DTMC (defined by a subset of the states and the transitions).
- A subsystem is **critical** iff all paths that lie inside the subsystem and lead to a target state form a counterexample.
- A critical subsystem is **state-minimal** iff it is a critical subsystem with the smallest number of states.
- Analogously: **transition-minimal**

Variables:

- for each $s \in S$ a variable $x_s \in [0, 1] \subseteq \mathbb{Z}$ encodes if s belongs to the subsystem or not, and
- for each $s \in S$ a variable $p_s \in [0, 1] \subseteq \mathbb{R}$ encodes the probability to reach a target state from s within the subsystem.

Minimal critical subsystems via SMT-solving

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min $\sum_{s \in S} x_s$ such that

- $p_{s_{init}} > \lambda$
- $\forall s \in T :$
 $((x_s = 0 \wedge p_s = 0) \oplus (x_s = 1 \wedge p_s = 1))$
- $\forall s \in S \setminus T :$
 $((x_s = 0 \wedge p_s = 0) \oplus (x_s = 1 \wedge p_s = \sum_{s' \in succ(s)} P(s, s') \cdot p_{s'}))$.

$\min\left(-\frac{1}{2}p_{s_{init}} + \sum_{s \in \mathcal{S}} x_s\right)$ such that

- $p_{s_{init}} > \lambda$
- $\forall s \in \mathcal{S} \cap \mathcal{T} : p_s = x_s$
- $\forall s \in \mathcal{S} \setminus \mathcal{T} : p_s \leq x_s \quad \wedge \quad p_s \leq \sum_{s' \in \text{succ}(s)} P(s, s') \cdot p_{s'}$

Model	without redundant constraints			optimal configuration		
	Z3	SCIP	CPLEX	Z3	SCIP	CPLEX
crowds2-2	0.45	0.01	0.02 (0.03)	0.18	0.01	0.01 (0.03)
crowds2-3	6.70	0.16	1.33 (0.28)	4.82	0.12	0.06 (0.11)
crowds2-4	127.18	0.47	0.30 (0.24)	23.72	0.30	0.30 (0.24)
crowds2-5		0.90	0.56 (0.45)	152.28	0.60	0.56 (0.24)
crowds3-3		0.64	0.49 (0.33)		0.35	0.38 (0.30)
crowds3-4		4.29	5.53 (2.07)		1.45	0.89 (0.58)
crowds3-5		23.49	6.66 (2.77)		5.58	1.51 (0.87)
crowds5-4		743.84	14.23 (5.07)		13.28	12.51 (4.89)
crowds5-6			302.03 (38.39)		1947.46	100.26 (23.52)
crowds5-8						1000.79 (145.84)
leader3-2	0.06	0.07	0.62 (0.22)	0.05	0.01	0.21 (0.13)
leader3-3		91.89	0.43 (0.22)		0.06	0.02 (0.06)
leader3-4		2346.59	0.70 (0.36)		0.40	0.07 (0.09)
leader3-5		0.44	1.11 (0.55)		0.05	0.14 (0.18)
leader4-2	3.57	0.23	0.45 (0.21)	1.38	0.07	0.24 (0.17)
leader4-3		1390.79	22.33 (3.38)		0.21	0.49 (0.37)
leader4-4					1.49	1.88 (1.21)
leader4-5					1.15	4.06 (2.80)
leader4-6						8.70 (5.92)