Resource-Parameterized Program Analysis using Observation Sequences

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Complicated shared-state interaction

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Subtle bugs, easily missed by humans

Complicated thread-local behavior

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Undecidable in combination with shared state

Complicated thread-local behavior



Undecidable in combination with shared state

Can decide under a fixed resource bound:

"No violations within 8 context switches"

Complicated thread-local behavior



Undecidable in combination with shared state

Can decide under a fixed resource bound:

"No violations within 8 context switches"

But what if violation would appear after 9?

New proof technique

Checking safety in concurrent systems

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Prove, not just refute

New proof technique

Checking safety in concurrent systems

Prove, not just refute

Practical automation despite undecidability

• Past work: Context-bounded analysis

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- The CUBA Algorithm

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$$(Q, \Sigma, \Delta, q^I)$$

- Q States (shared)
- ∑ Stack alphabet
- Δ Program: relates $(Q \times \Sigma^n)$ and $(Q \times \Sigma^{n\pm 1})$
- $q^l \in Q$ Starting state

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Nondeterminism models acting on input

Concurrent pushdown system (CPDS)

Machine: Fixed collection of PDSes

$$\mathcal{P}_i = (Q, \Sigma_i, \Delta_i, q^I)$$

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Machine: Fixed collection of PDSes

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Shared: state set, current state

Per-PDS: stack alphabet, program

Decidability

✓ Control stack only

Decidability

- ✓ Control stack only
- ♦ Shared state only

Decidability

- ✓ Control stack only
- ✓ Shared state only
- Both at once

Context-bounded anaysis (CBA)

Finite bound on context switches

Context-bounded analysis (CBA)

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Can we violate safety property with only 5 context switches?

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Finitely many single-PDS reachability questions

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Finitely many single-PDS reachability questions

Can only refute safety properties, not prove them

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Context-unbounded analysis (CUBA)

Proof technique built on CBA

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CBA: "No safety violation within k context switches"

Context-unbounded analysis (CUBA)

Proof technique built on CBA

CBA: "No safety violation within k context switches"

CUBA: "No k is big enough for CBA to observe safety violation"

Observation Sequence

Definition parameterized over

- \mathcal{P} a CPDS (program we're investigating)
- \mathcal{D} a poset (space of possible observations)
- C a property of resource-bounded CPDSes (what we're trying to check)

Observation Sequence

An observation sequence for property C on machine P over a poset D is a sequence $(O_k)_{k=0}^{\infty}$ with the following properties

Monotonicity: $\forall k \in \mathbb{N}.O_k \sqsubseteq O_{k+1}$

Computability: There is a computable function $f: \mathbb{N} \to \mathcal{D}$ such that $f(k) = O_k$

Expressibility: There is a computable predicate p on $\{O_k|k\in\mathcal{N}\}$ such that $p(O_k)$ holds iff \mathcal{P} has property \mathcal{C} when subject to resource bound k

Why observation sequences?

Definition captures important properties for safety proofs

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Definition captures important properties for safety proofs

Monotonicity: With more resources, more results are possible

Computability: We can actually make these observations

Expressibility: What we observe informs us about what we care about

Guaranteeing convergence

Unbounded stack size
Infinite state space

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Observe only top of stack Finite observation poset

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Monotonic sequence must eventually converge

Monotone data flow analysis?

Data flow analysis: iterate function to find least fixed point

$$f:D\to D$$

Monotone data flow analysis?

Data flow analysis: iterate function to find least fixed point

$$f:D\to D$$

CUBA: grow input until convergence

$$f: \mathbb{N} \to D$$

Monotone data flow analysis?

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CUBA: grow input until convergence

$$f: \mathbb{N} \to D$$

No fixed points!

$$f(n+1) = f(n) \not\Rightarrow f(n+2) = f(n)$$

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 Are we done yet? We might be...

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How do we know when to stop?

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Recall: CPDS reachability is undecidable!

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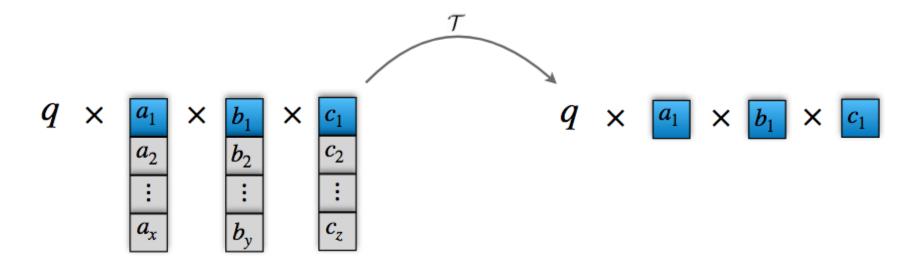
Recall: CPDS reachability is undecidable!

Key automation challenge: distinguish stuttering from convergence

Contents

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Finitize the observation sequences to guarantee convergence



Stutter Detection is undecidable

Since convergence is guaranteed, but the problem is undecidable, stutter detection is undecidable

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Must approximate!

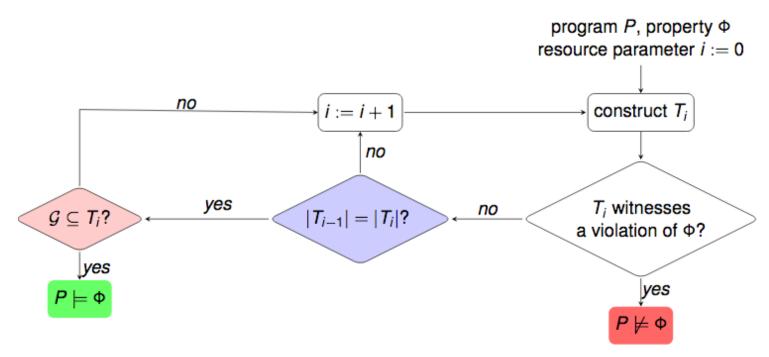
Only certain instructions can change observation set

If a stutter ends it has to be the consequence of a stack pop.

The observation sequence converges if all reachable popping states are in the observation set

We give an overapproximation of this set to stay decidable

Algorithm Summary



 $G := \{ reachable generators \}$

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Empirical Evaluation

Evaluated our CUBA algorithm on 9 CPDSs converted from concurrent programs in C/Java

For comparison, evaluated most of those using JMoped, a CBA implementation

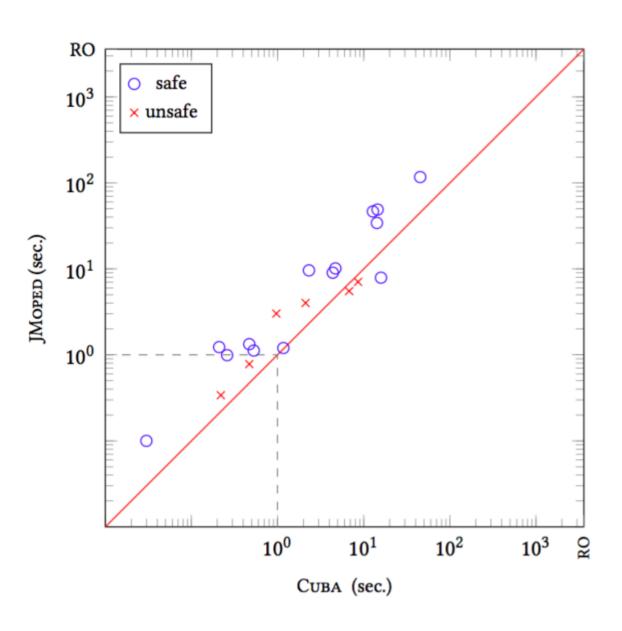
CUBA is effective at proving and refuting safety

ID/Program	Prog. Feat	$(R_k)_{k=0}^{\infty}$	$(\mathcal{T}(R_k))_{k=0}^{\infty}$			
	Thread FCR?	Safe?	k_{max}	k_{max}	Time	Mem
1/Вьшетоотн-1	1+1	Х	≥ 7	6 (4)	0.26	18.14
	1 + 2	X	≥ 7	6 (3)	2.32	136.26
	2 + 1	X	≥ 8	7 (4)	12.76	347.74
2/Вьшетоотн-2	1+1	X	≥ 7	6 (4)	0.53	23.43
	1 + 2	X	≥ 7	6 (3)	4.39	196.73
	2 + 1	X	≥ 8	7 (4)	14.21	387.23
3/Вьшетоотн-3	1+1	✓	≥ 7	6	0.47	22.15
	1 + 2	✓	≥ 7	6	4.71	180.11
	2 + 1	/	≥ 8	7	14.46	375.42

ID/Program	Prog. Features			$(R_k)_{k=0}^{\infty}$	$(\mathcal{T}(R_k))_{k=0}^{\infty}$		
	Thread	FCR?	Safe?	k_{max}	k_{max}	Time	Mem
4/BST-Insert	1+1	•	/	2	2	1.17	24.53
	2 + 1	•	✓	3	3	15.84	140.93
	2 + 2	•	✓	≥ 5	4	45.21	355.74
5/FileCrawler	1° + 2	•	✓	6	6	0.03	5.35
6/K-Induction	1+1	0	/	≥ 4	3	0.23	3.78
7/Proc-2	2 + 2*	0	✓	≥ 4	3	0.52	18.04
8/STEFAN-1	2	0	/	≥ 3	2	1.01	2.81
	4	0	✓	≥ 5	4	16.36	1185.62
	8	0	-	≥ 8	≥ 8	-	OOM
9/Dekker	2*	•	/	6	6	0.21	13.42

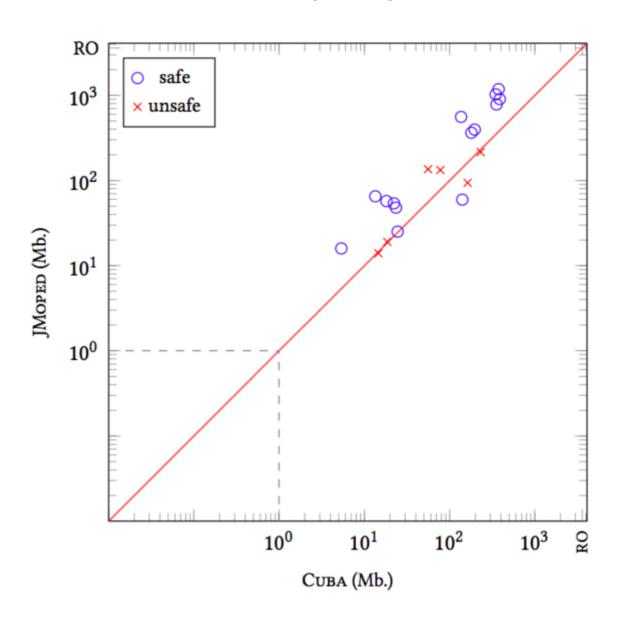
CUBA has competitive performance with CBA

Runtime



CUBA has competitive performance with CBA

Memory Usage



More in the paper!

- Systematic theory of observation sequences
- How we represent obvervation sequences efficiently?
- How to compute generator sets

Conclusions

- Context UnBounded Analysis can automatically find bugs and prove safety of concurrent programs
- Observation sequences provide unifying perspective for understanding resource constrained analyses
- As efficient as previous, weaker analyses