

Self-Adaptive Static Analysis

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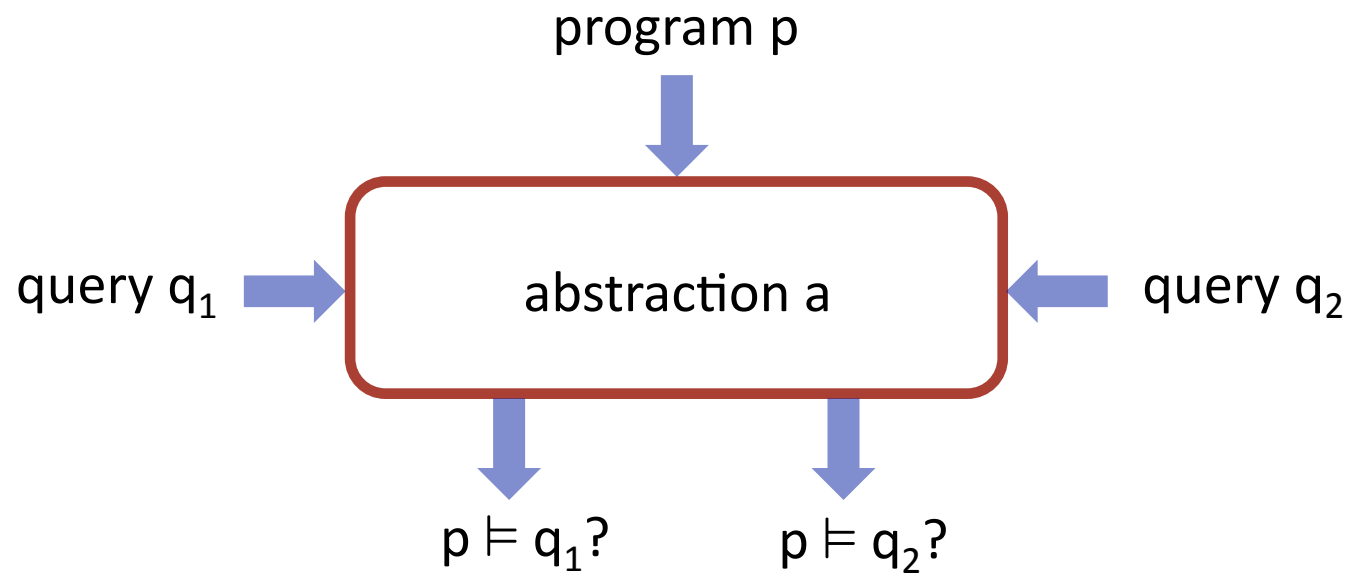
Stanford University

Static Analysis: 70's to 90's

- ▶ client-oblivious

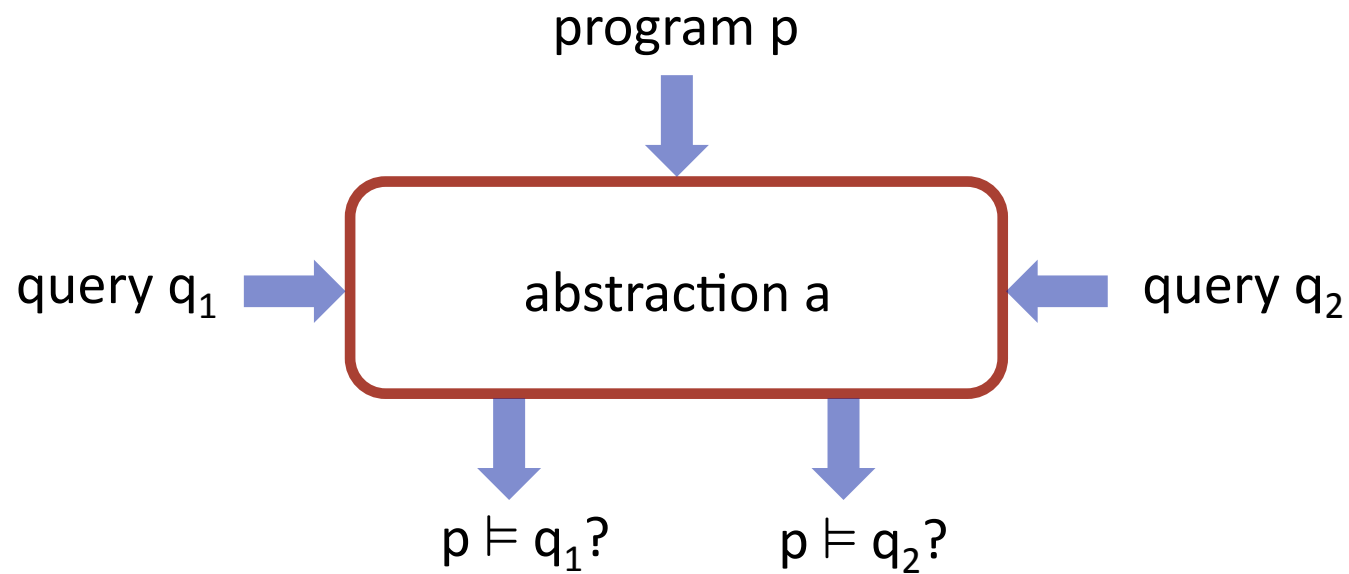
“Because clients have different precision and scalability needs, future work should identify the client they are addressing ...”

M. Hind, *Pointer Analysis: Haven't We Solved This Problem Yet?*, 2001



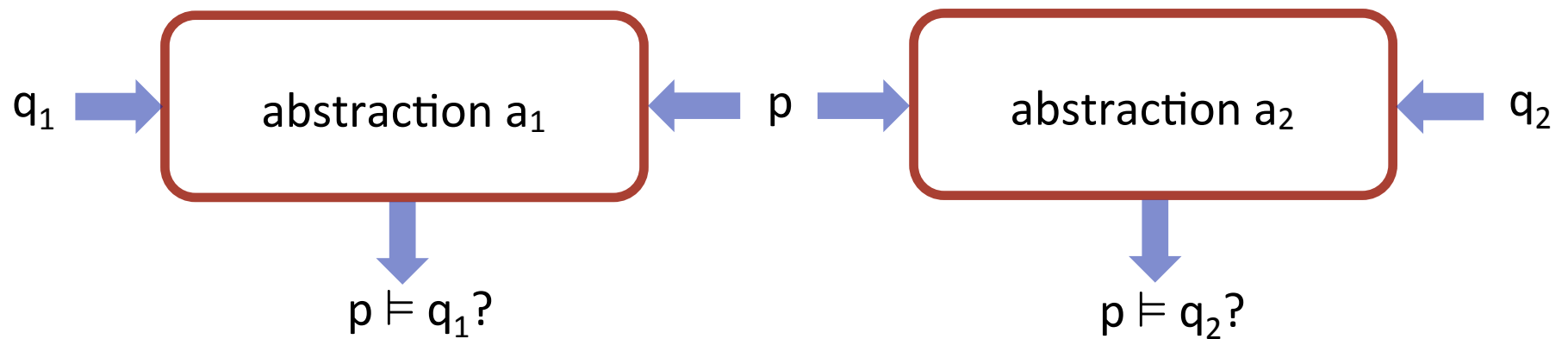
Static Analysis: 00's to Present

- ▶ client-driven



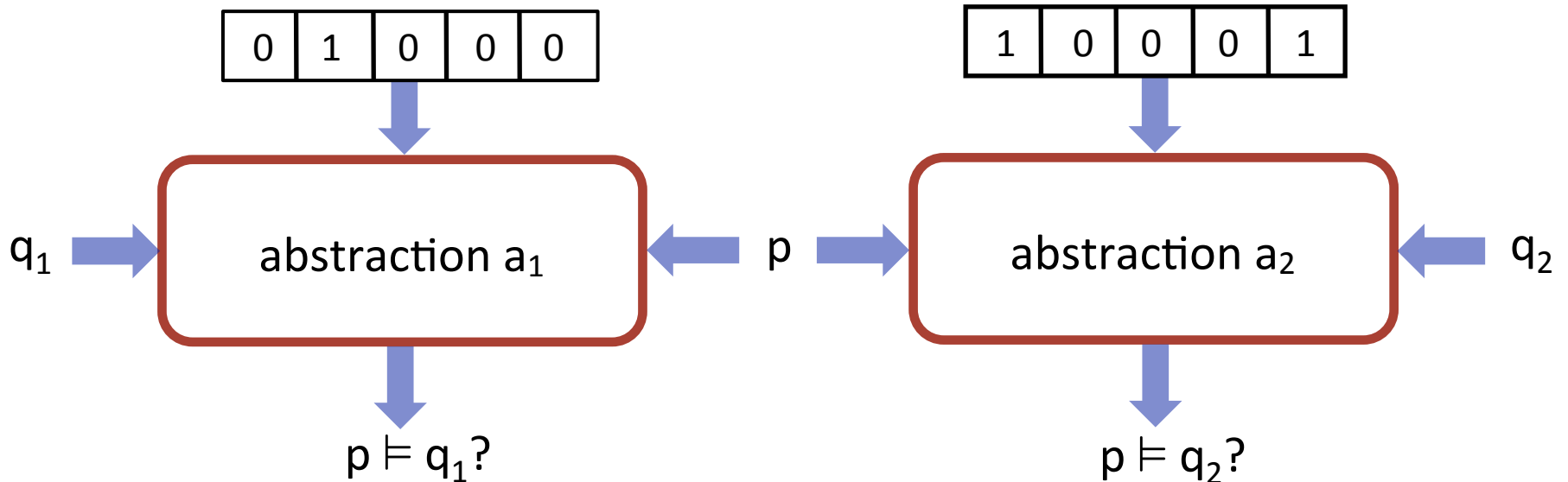
Static Analysis: 00's to Present

- ▶ client-driven
 - ▶ modern pointer analyses
 - ▶ software model checkers



Our Static Analysis Setting





- ▶ client-driven + parametric
 - ▶ new search algorithms: testing, machine learning, ...
 - ▶ new analysis questions: optimality, impossibility, ...



Example 1: Type-State Analysis

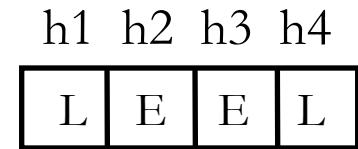
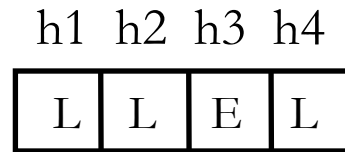
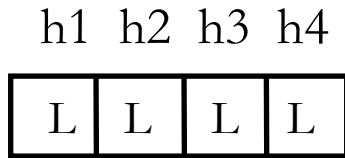
Must-Alias Set to Track:

```
x = new File;  
y = x;  
if (*) z = x;  
x.open();  
y.close();  
if (*)  
    assert1(x, closed);  
else  
    assert2(x, opened);
```

$\{x\}$	$\{x, y\}$
$\langle closed, \{x\} \rangle$	$\langle closed, \{x\} \rangle$
$\langle closed, \{x\} \rangle$	$\langle closed, \{x, y\} \rangle$
$\langle closed, \{x\} \rangle$	$\langle closed, \{x, y\} \rangle$
$\langle opened, \{x\} \rangle$	$\langle opened, \{x, y\} \rangle$
$\langle closed \vee opened, \{x\} \rangle$	$\langle closed, \{x, y\} \rangle$
	
	

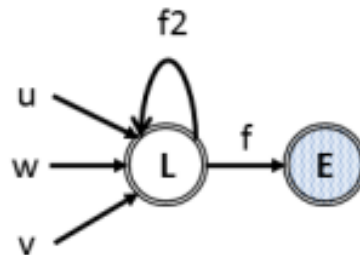
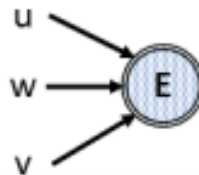
Example 2: Thread-Escape Analysis

Heap abstraction:

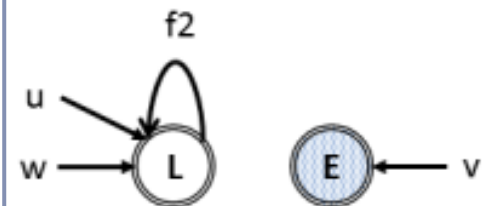


```

while (*) {
  u = new h1;
  v = new h2;
  g = new h3;
  v.f = g;
  w = new h4;
  u.f2 = w;
  assert(local(w));
  u.spawn();
}
    
```






but not optimal



and optimal!

Example 3: Approximation Safety Analysis

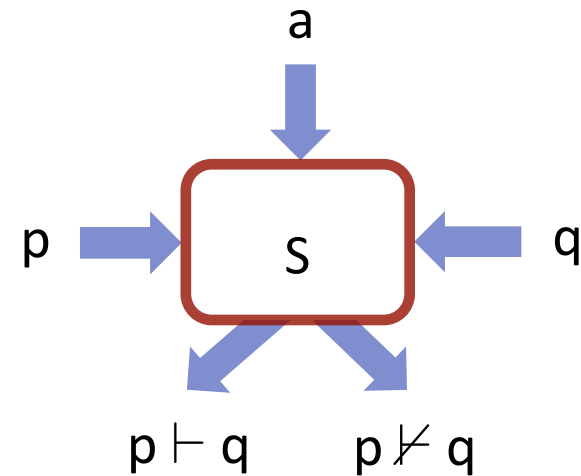
Approximated Operations:	{ 1, 2, 3, 4 }	{ 2, 3, 4 }	{ 2, 4 }
1: v1 = input();	{ {v1} }	{ ∅ }	{ ∅ }
2: v2 = input();	{ {v1,v2} }	{ {v2} }	{ {v2} }
restrict(v1);	{ T }	{ {v2} }	{ {v2} }
while (v1 > 0) {	{ T }	{ {v2}, T }	{ {v2} }
3: v1 = f(v1);	{ T }	{ {v1,v2}, T }	{ {v2} }
4: v2 = g(v2);	{ T }	{ {v1,v2}, T }	{ {v2} }
restrict(v1);	{ T }	{ T }	{ {v2} }
}	{ T }	{ {v2}, T }	{ {v2} }
relax(v2);	{ T }	{ ∅, T }	{ ∅ }
restrict(v2);	{ T }	{ ∅, T }	{ ∅ }
output(v2);	{ T }	{ ∅, T }	{ ∅ }
			

Problem Statement

- ▶ An efficient algorithm with:

INPUTS:

- ▶ program p and query q
- ▶ abstractions $A = \{ a_1, \dots, a_n \}$
- ▶ boolean function $S(p, q, a)$



OUTPUT:

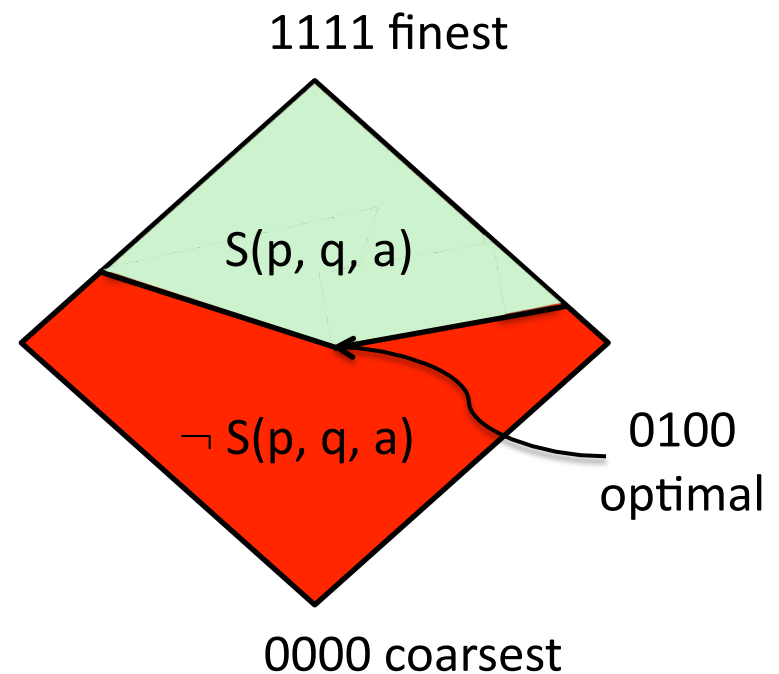
- ▶ Impossibility: $\nexists a \in A: S(p, q, a) = \text{true}$
- ▶ or Proof: some $a \in A: S(p, q, a) = \text{true}$ **AND**

$$\forall a' \in A: (a' \leq a \wedge S(p, q, a') = \text{true}) \Rightarrow a' = a$$

Optimal Abstraction

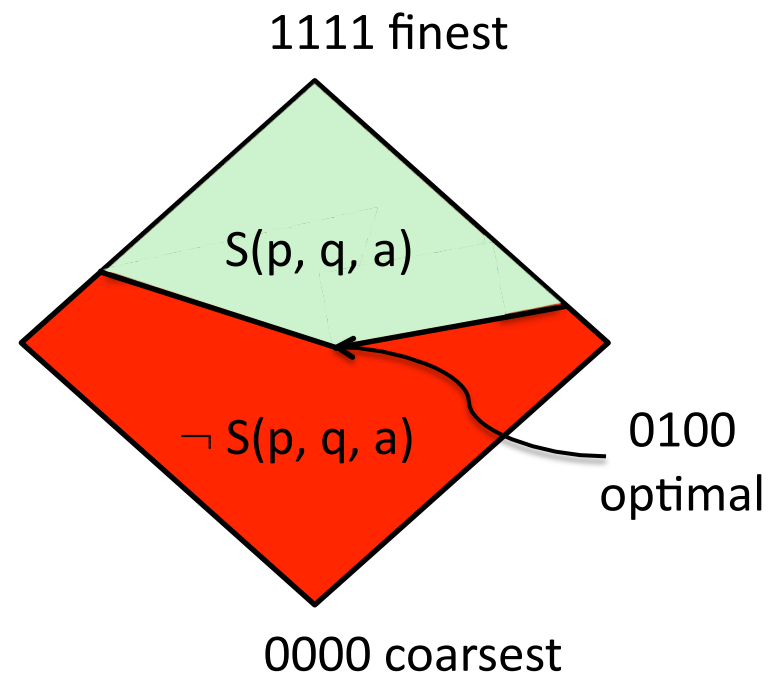
Why Optimal Abstraction?

- ▶ Yields smallest/largest solution
- ▶ Provides empirical lower bounds
- ▶ Efficient to compute
- ▶ Better for user consumption
 - ▶ analysis imprecision facts
 - ▶ assumptions about missing program parts
- ▶ Suitable for machine learning



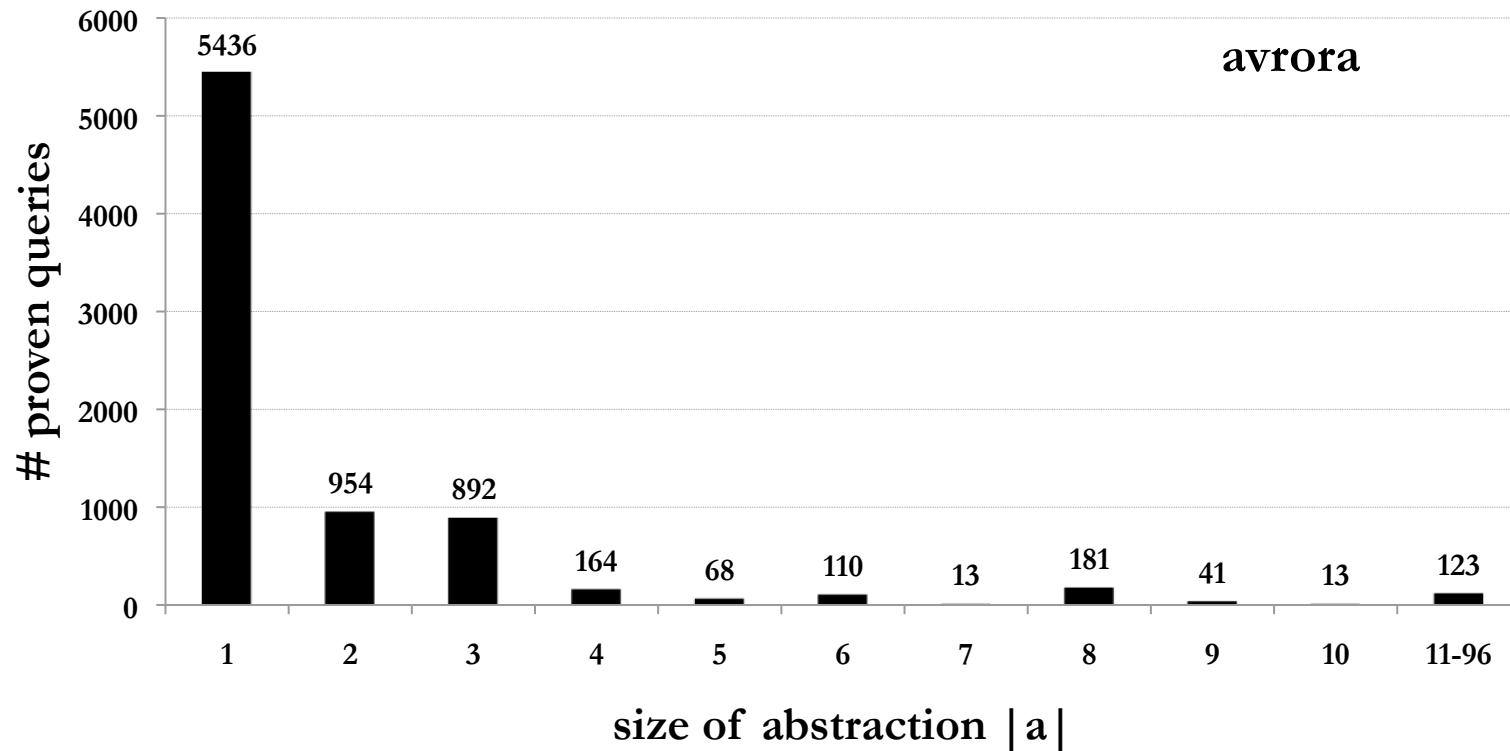
Why is this Hard in Practice?

- ▶ $|A|$ exponential in size of \mathbf{p} , or even infinite
- ▶ $S(\mathbf{p}, \mathbf{q}, \mathbf{a}) = \text{false}$ for most $\mathbf{p}, \mathbf{q}, \mathbf{a}$
- ▶ Different \mathbf{a} is optimal for different \mathbf{p}, \mathbf{q}

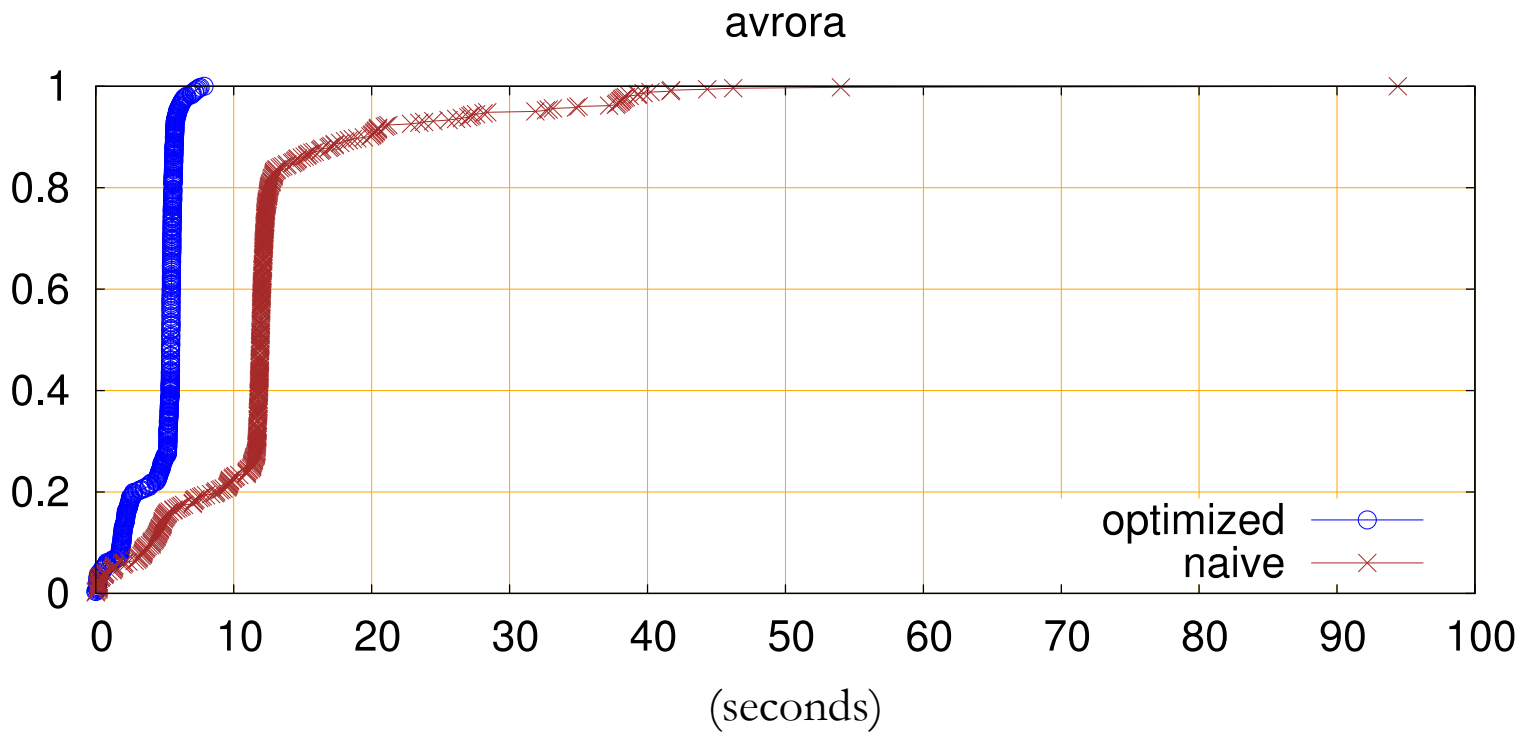


Example: Size of Optimal Abstractions

$|A| = 2^{10k}$ $|Q| = 14k$ $|Q_{\text{proven}}| = 55\%$ of $|Q|$



Example: Runtime under Optimal Abstractions



Summary of Our Results

- ▶ Machine Learning [POPL'11]
 - ▶ “Learning Minimal Abstractions”
- ▶ Dynamic Analysis [POPL'12]
 - ▶ “Abstractions from Tests”
- ▶ Static Refinement [PLDI'13]
 - ▶ “Finding Optimum Abstractions in Parametric Dataflow Analysis”
- ▶ Constraint Solving [PLDI'14]
 - ▶ “On Abstraction Refinement for Program Analyses in Datalog”

All implementations available in Chord, an extensible program analysis framework for Java bytecode (jchord.googlecode.com).

Datalog for Program Analysis



Soot



DOOP



What is Datalog?



What is Datalog?



Input relations: $\text{edge}(i, j)$.

Output relations: $\text{path}(i, j)$.

Rules: (1) $\text{path}(i, i)$.

(2) $\text{path}(i, k) \text{ :- path}(i, j), \text{edge}(j, k)$.

Least fixpoint computation:

Input: $\text{edge}(0, 1), \text{edge}(1, 2)$.

$\text{path}(0, 0)$.

$\text{path}(1, 1)$.

$\text{path}(2, 2)$.

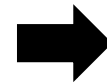
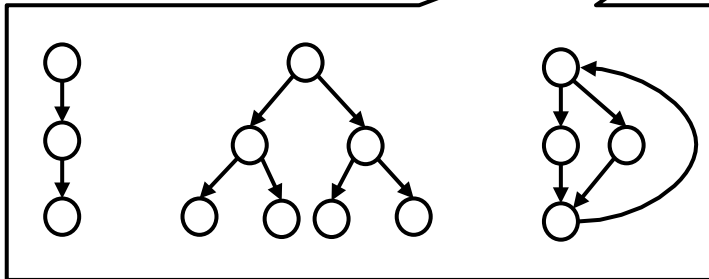
$\text{path}(0, 1) \text{ :- path}(0, 0), \text{edge}(0, 1)$.

$\text{path}(0, 2) \text{ :- path}(0, 1), \text{edge}(1, 2)$.

Why Datalog?

If there exists a path from **a** to **b**, and there is an edge from **b** to **c**, then there exists a path from **a** to **c**:

$\text{path}(\mathbf{a}, \mathbf{c}) \text{ :- path}(\mathbf{a}, \mathbf{b}), \text{edge}(\mathbf{b}, \mathbf{c}).$



Why Datalog?



k-object-sensitivity,
 $k = 2, \sim 100\text{KLOC}$



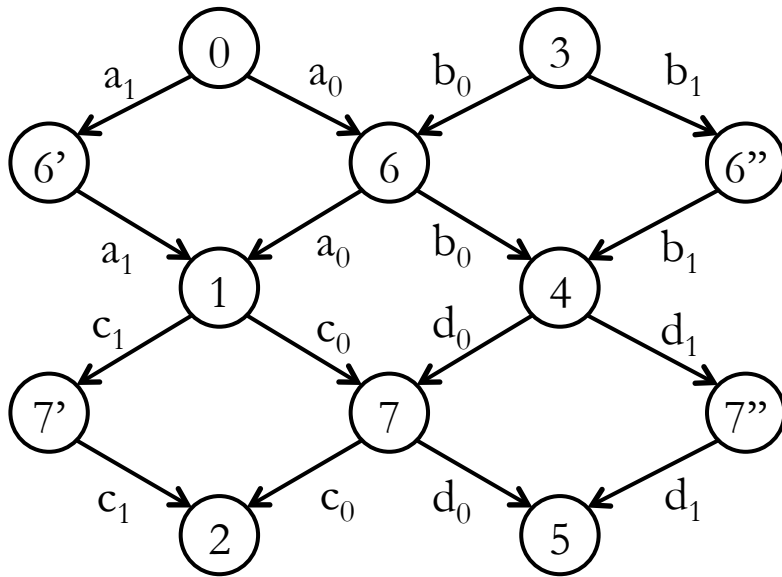
Limitation

k-object-sensitivity,
 $k = 2$, ~100KLOC



k-object-sensitivity,
 $k = 10$, ~500KLOC

Pointer Analysis as Graph Reachability



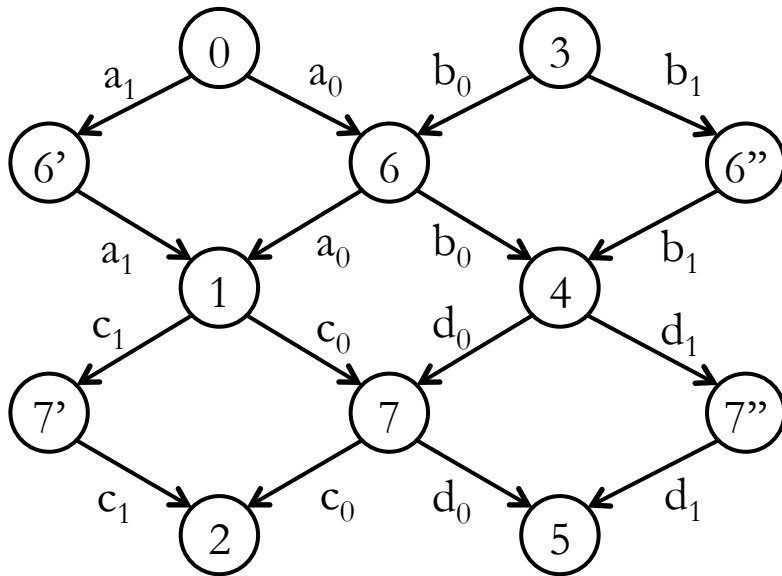
```
f() {  
    v1 = new ...;  
    v2 = id1(v1);  
    v3 = id2(v2);  
    q2: assert(v3 != v1);  
}
```

```
g() {  
    v4 = new ...;  
    v5 = id1(v4);  
    v6 = id2(v5);  
    q1: assert(v6 != v1);  
}
```

```
id1(v) {return v;}
```

```
id2(v) {return v;}
```

Graph Reachability in Datalog



Input relations:

$\text{edge}(i, j, n), \text{abs}(n)$

Output relations:

$\text{path}(i, j)$

Rules:

$\text{path}(i, i).$

$\text{path}(i, j) :- \text{path}(i, k), \text{edge}(k, j, n), \text{abs}(n).$

Query Tuple	Original Query
$q_1: \text{path}(0, 5)$	assert ($v_6 \neq v_1$)
$q_2: \text{path}(0, 2)$	assert ($v_3 \neq v_1$)

Input tuples:

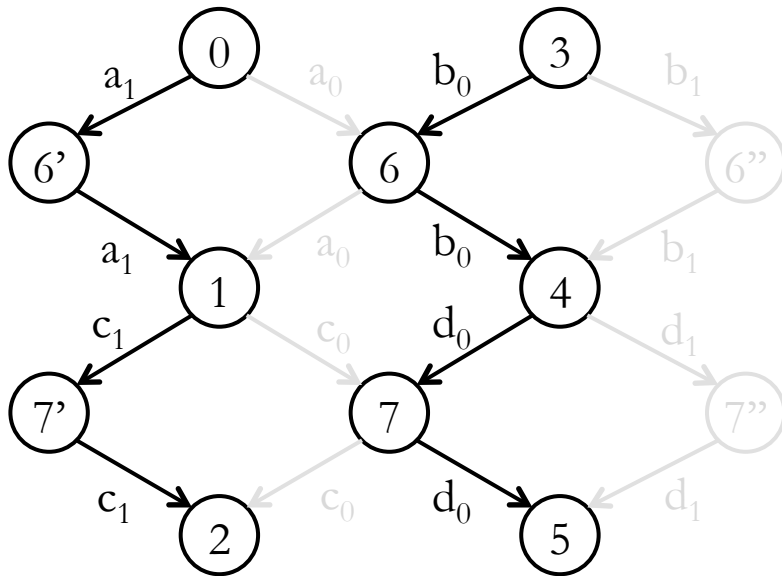
$\text{edge}(0, 6, a_0), \text{edge}(0, 6', a_1), \text{edge}(3, 6, b_0),$

...

$\text{abs}(a_0) \oplus \text{abs}(a_1), \text{abs}(b_0) \oplus \text{abs}(b_1),$
 $\text{abs}(c_0) \oplus \text{abs}(c_1), \text{abs}(d_0) \oplus \text{abs}(d_1).$

16 possible abstractions in total

Desired Result



Input relations:

$\text{edge}(i, j, n), \text{abs}(n)$

Output relations:

$\text{path}(i, j)$

Rules:

$\text{path}(i, i).$

$\text{path}(i, j) :- \text{path}(i, k), \text{edge}(k, j, n), \text{abs}(n).$

Input tuples:

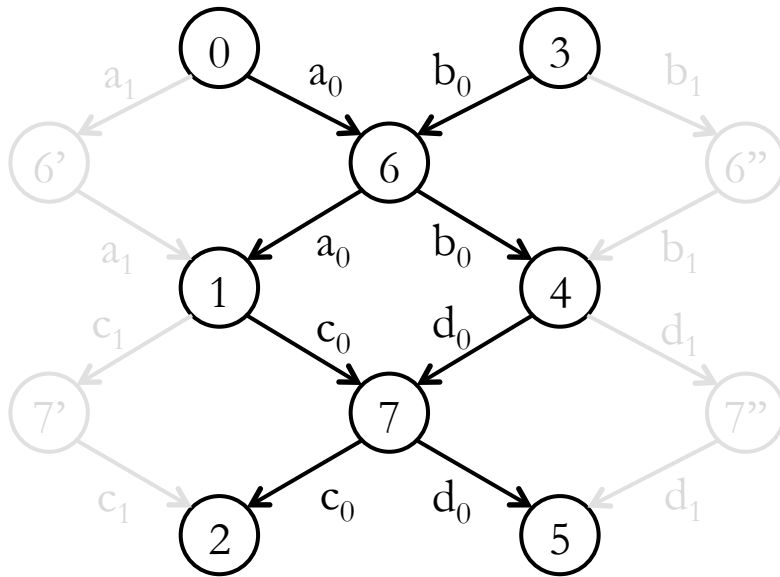
$\text{edge}(0, 6, a_0), \text{edge}(0, 6', a_1), \text{edge}(3, 6, b_0),$

...

$\text{abs}(a_0) \oplus \text{abs}(a_1), \text{abs}(b_0) \oplus \text{abs}(b_1),$
 $\text{abs}(c_0) \oplus \text{abs}(c_1), \text{abs}(d_0) \oplus \text{abs}(d_1).$

Query	Answer
$q_1: \text{path}(0, 5)$	✓ $a_1 b_0 c_1 d_0$
$q_2: \text{path}(0, 2)$	✗ Impossibility

Iteration 1



```

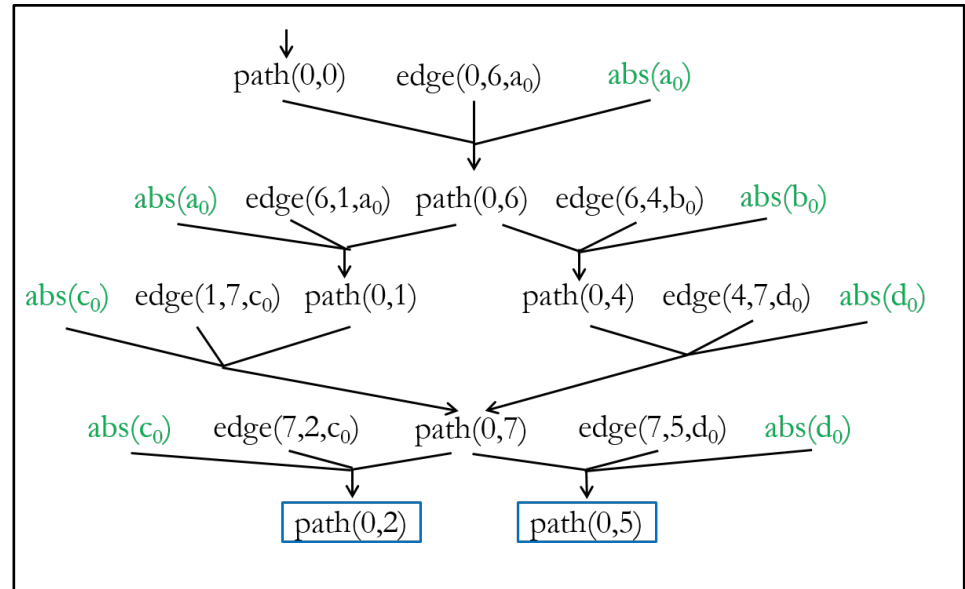
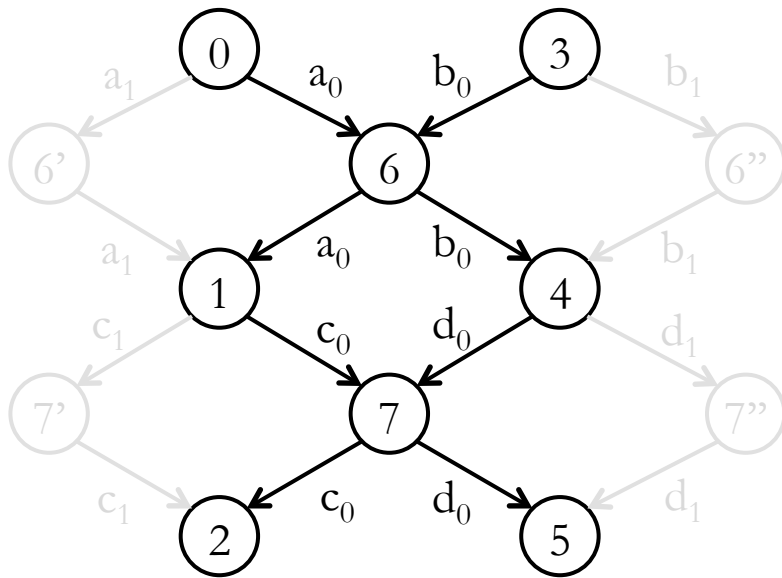
path(0, 0).
path(0, 6) :- path(0, 0), edge(0, 6, a0), abs(a0).
path(0, 1) :- path(0, 6), edge(6, 1, a0), abs(a0).
path(0, 7) :- path(0, 1), edge(1, 7, c0), abs(c0).
path(0, 2) :- path(0, 7), edge(7, 2, c0), abs(c0).
path(0, 4) :- path(0, 6), edge(6, 4, b0), abs(b0).
path(0, 7) :- path(0, 4), edge(4, 7, d0), abs(d0).
path(0, 5) :- path(0, 7), edge(7, 5, d0), abs(d0).
...

```

Query	Eliminated Abstractions
q_1 : path(0, 5)	
q_2 : path(0, 2)	

$$\text{abs}(a_0) \oplus \text{abs}(a_1), \text{abs}(b_0) \oplus \text{abs}(b_1), \\
 \text{abs}(c_0) \oplus \text{abs}(c_1), \text{abs}(d_0) \oplus \text{abs}(d_1).$$

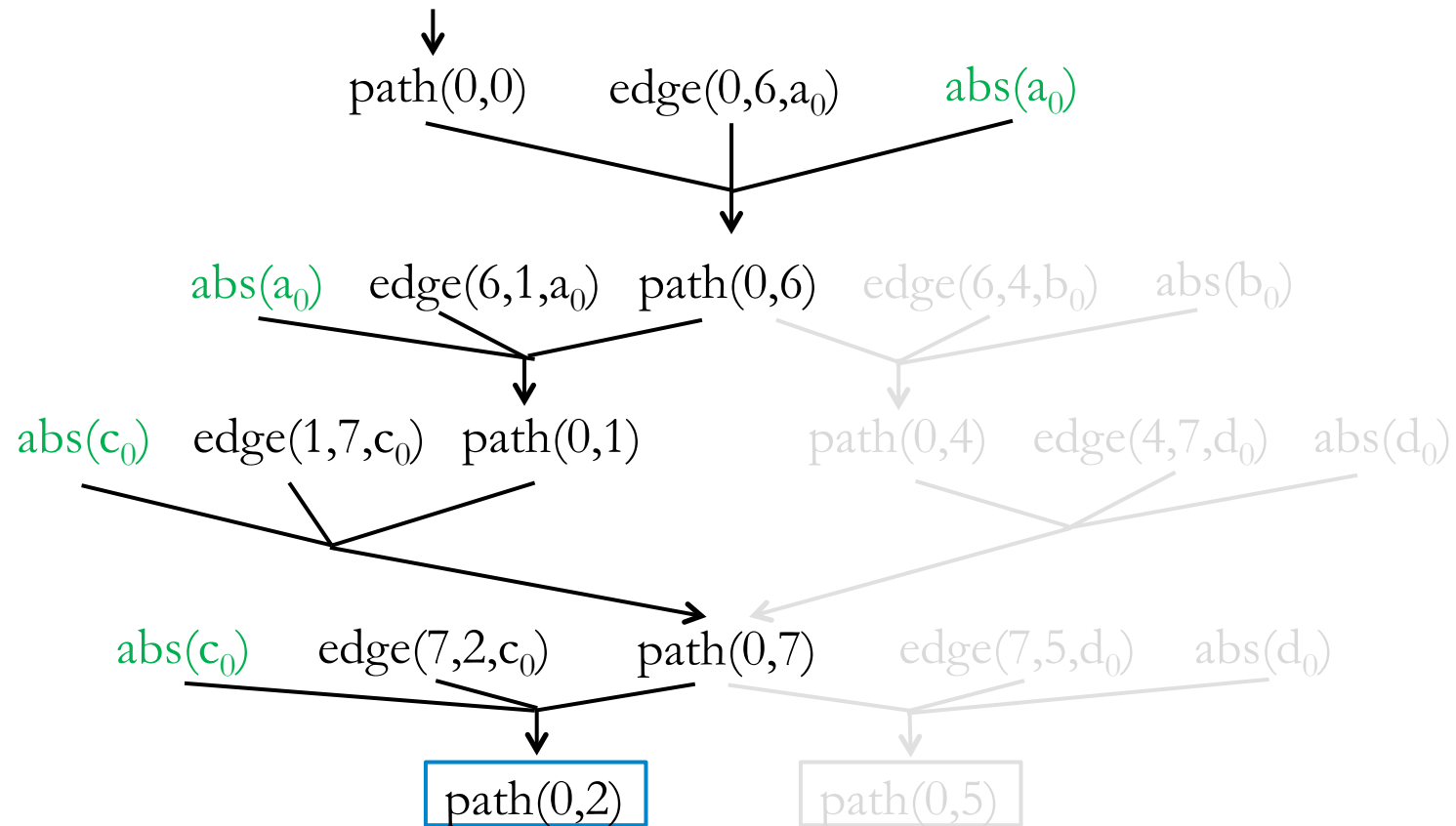
Iteration 1 - Derivation Graph



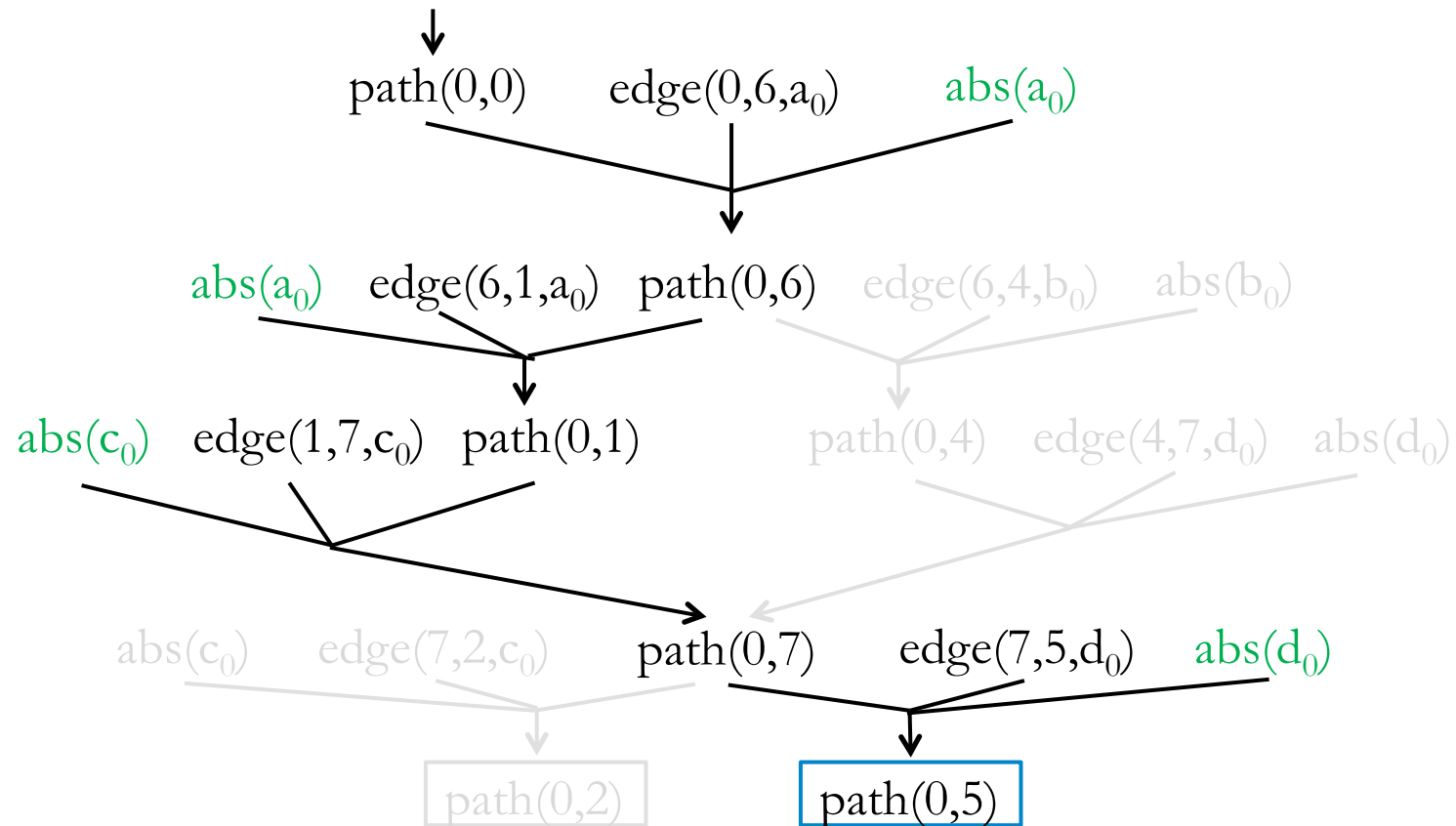
Query	Eliminated Abstractions
q_1 : path(0, 5)	
q_2 : path(0, 2)	

$$\begin{aligned}
 & \text{abs}(a_0) \oplus \text{abs}(a_1), \text{abs}(b_0) \oplus \text{abs}(b_1), \\
 & \text{abs}(c_0) \oplus \text{abs}(c_1), \text{abs}(d_0) \oplus \text{abs}(d_1).
 \end{aligned}$$

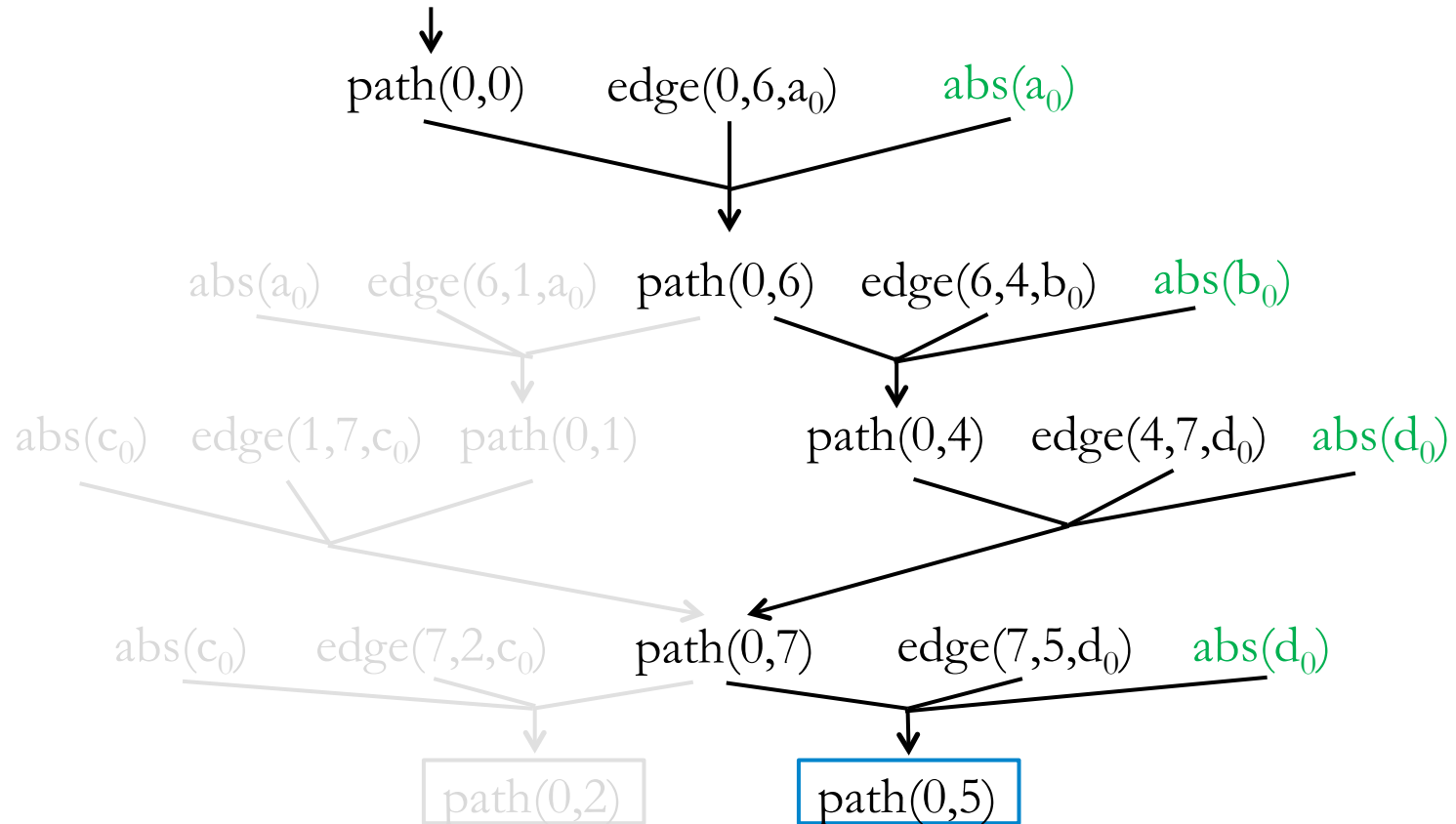
Iteration 1 - Derivation Graph



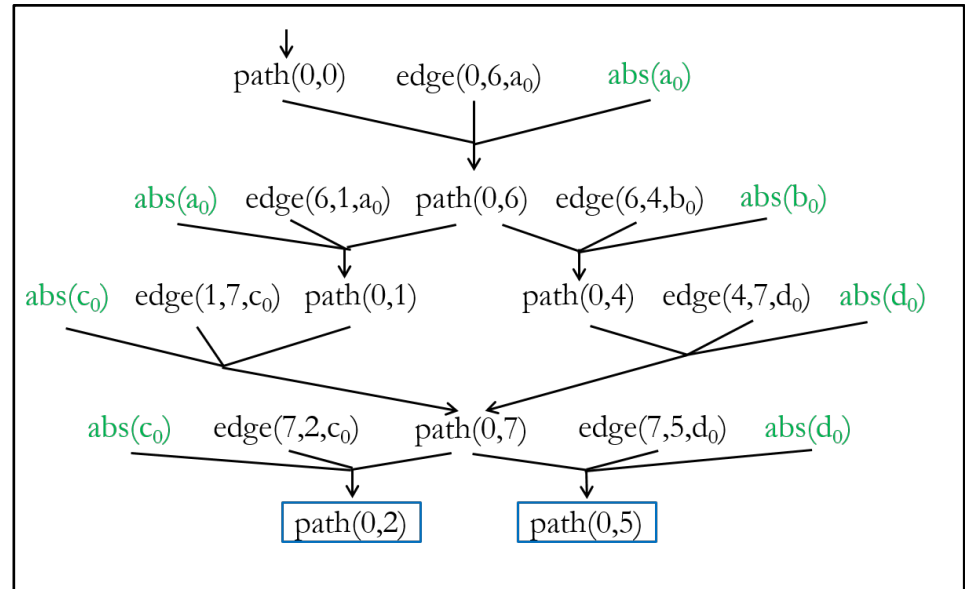
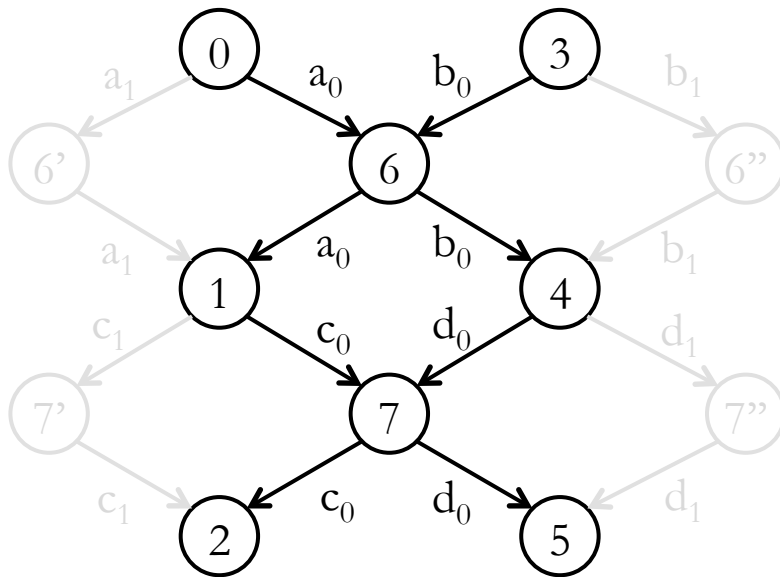
Iteration 1 - Derivation Graph



Iteration 1 - Derivation Graph



Iteration 1 - Derivation Graph



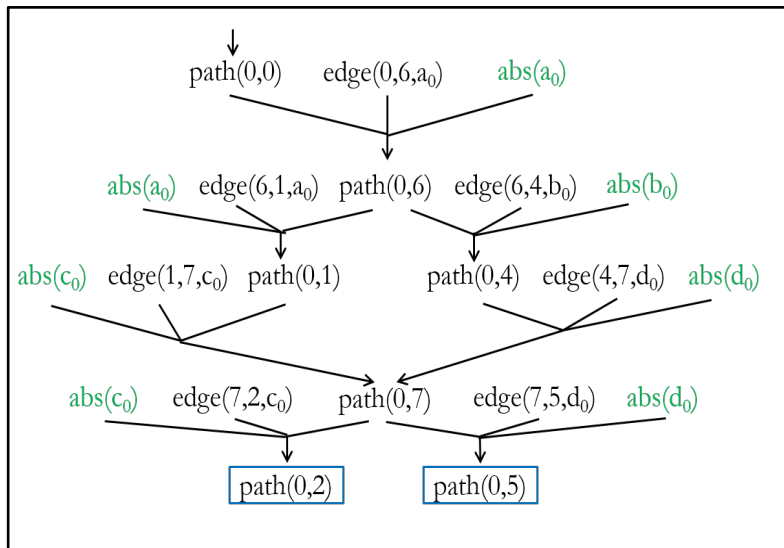
Query	Eliminated Abstractions
$q_1: \text{path}(0, 5)$	$a_0c_0d_0, a_0b_0d_0$ (4/16)
$q_2: \text{path}(0, 2)$	a_0c_0 (4/16)

$$\begin{aligned}
 & \text{abs}(a_0) \oplus \text{abs}(a_1), \text{abs}(b_0) \oplus \text{abs}(b_1), \\
 & \text{abs}(c_0) \oplus \text{abs}(c_1), \text{abs}(d_0) \oplus \text{abs}(d_1).
 \end{aligned}$$

Encoded as MAXSAT

Hard Constraints

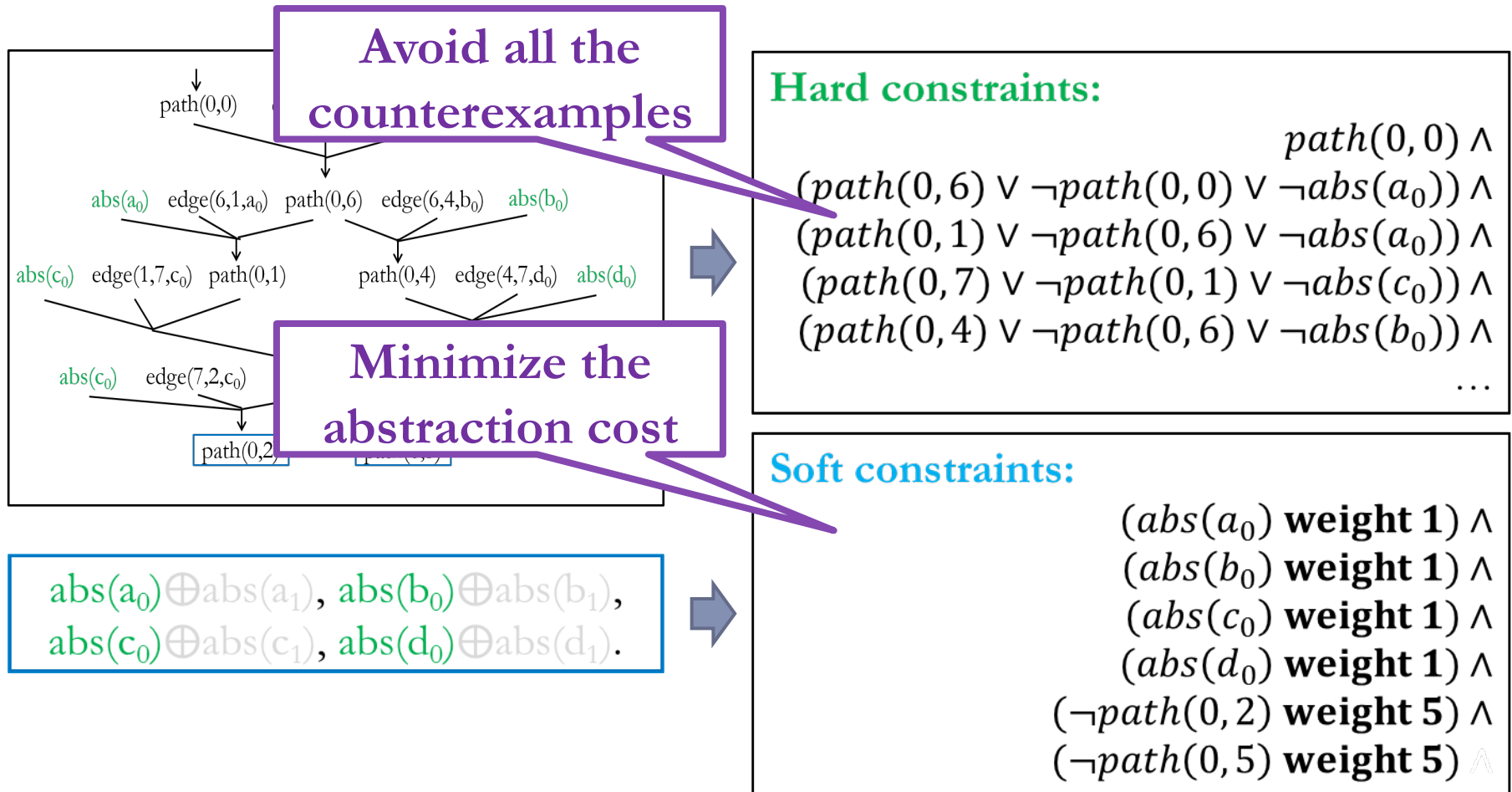
Soft Constraints



MAXSAT($\psi_0, (\psi_1, w_1), \dots, (\psi_n, w_n)$):

Find solution S that
 Maximize $\{\psi_i \mid S \models \psi_i, 1 \leq i \leq n\}$
 Subject to $S \models \psi_0$

Encoded as MAXSAT



Encoded as MAXSAT

Solution:

$path(0, 0) = \text{true}, path(0, 6) = \text{false},$
 $path(0, 1) = \text{false}, path(0, 4) = \text{false},$
 $path(0, 7) = \text{false}, path(0, 2) = \text{false},$
 $path(0, 5) = \text{false}, path(0, 3) = \text{true},$
 $abs(a_0) = \text{false}, abs(b_0) = \text{true},$
 $abs(c_0) = \text{true}, abs(d_0) = \text{true}.$


 $a_1 b_0 c_0 d_0$

Query	Eliminated Abstractions
$q_1: path(0, 5)$	$a_0 c_0 d_0, a_0 b_0 d_0$ (4/16)
$q_2: path(0, 2)$	$a_0 c_0$ (4/16)

Hard constraints:

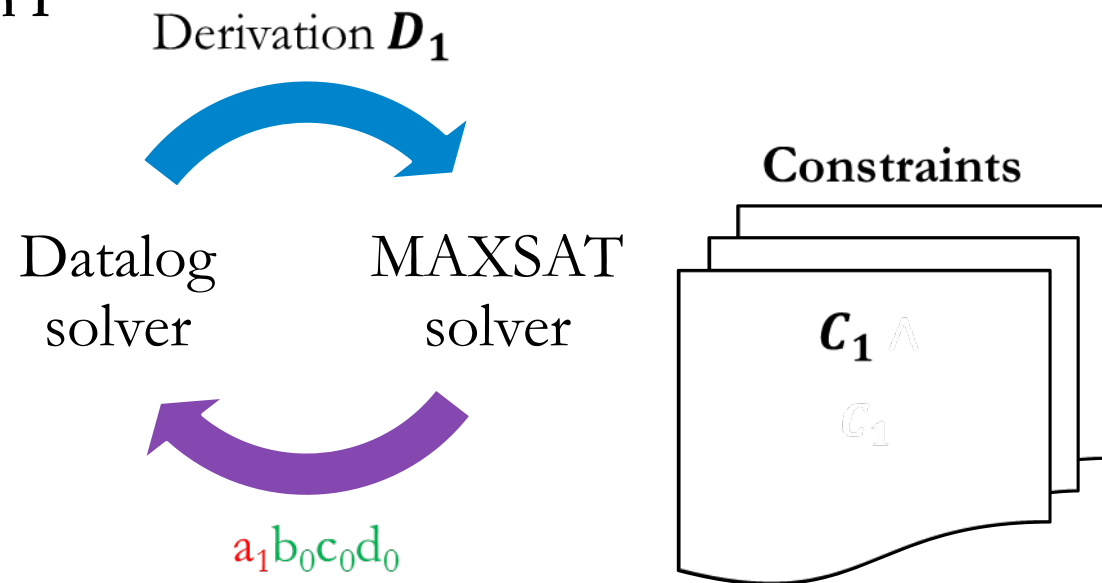
$path(0, 0) \wedge$
 $(path(0, 6) \vee \neg path(0, 0) \vee \neg abs(a_0)) \wedge$
 $(path(0, 1) \vee \neg path(0, 6) \vee \neg abs(a_0)) \wedge$
 $(path(0, 7) \vee \neg path(0, 1) \vee \neg abs(c_0)) \wedge$
 $(path(0, 4) \vee \neg path(0, 6) \vee \neg abs(b_0)) \wedge$
 ...

Soft constraints:

$(abs(a_0) \text{ weight } 1) \wedge$
 $(abs(b_0) \text{ weight } 1) \wedge$
 $(abs(c_0) \text{ weight } 1) \wedge$
 $(abs(d_0) \text{ weight } 1) \wedge$
 $(\neg path(0, 2) \text{ weight } 5) \wedge$
 $(\neg path(0, 5) \text{ weight } 5) \wedge$

Iteration 2 and Beyond

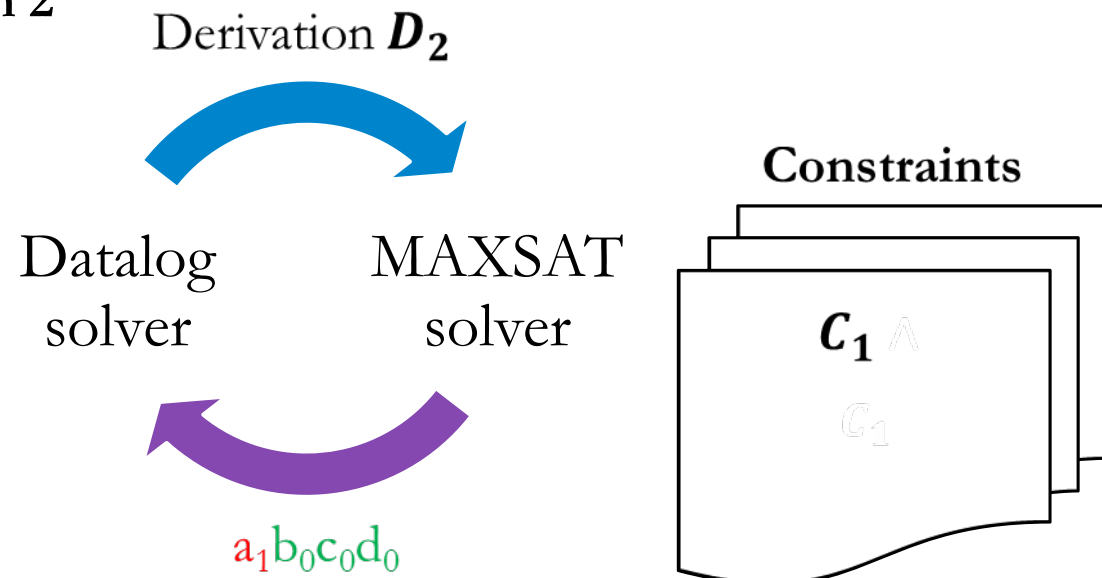
Iteration 1



Query	Answer	Eliminated Abstractions
q_1 : path(0, 5)		$a_0c_0d_0, a_0b_0d_0, a_1d_0$ (4/16)
q_2 : path(0, 2)		a_0c_0, a_1c_0 (4/16)

Iteration 2 and Beyond

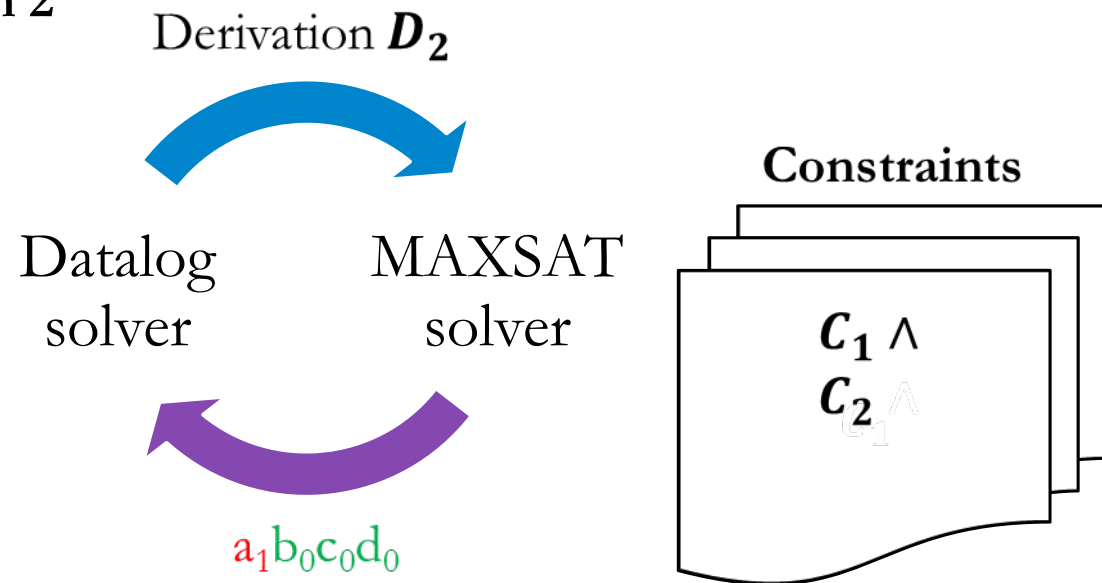
Iteration 2



Query	Answer	Eliminated Abstractions
q_1 : path(0, 5)		$a_0c_0d_0, a_0b_0d_0, a_1d_0$ (4/16)
q_2 : path(0, 2)		a_0c_0, a_1c_0 (4/16)

Iteration 2 and Beyond

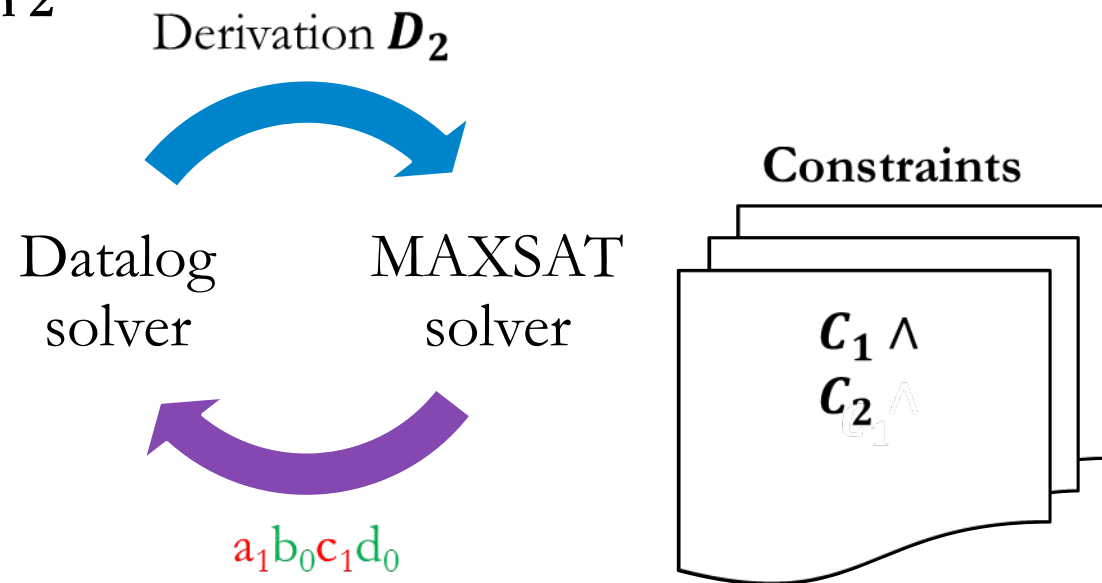
Iteration 2



Query	Answer	Eliminated Abstractions
q_1 : path(0, 5)		$a_0 c_0 d_0, a_0 b_0 d_0, a_1 d_0$ (4/16)
q_2 : path(0, 2)		$a_0 c_0$ (4/16)

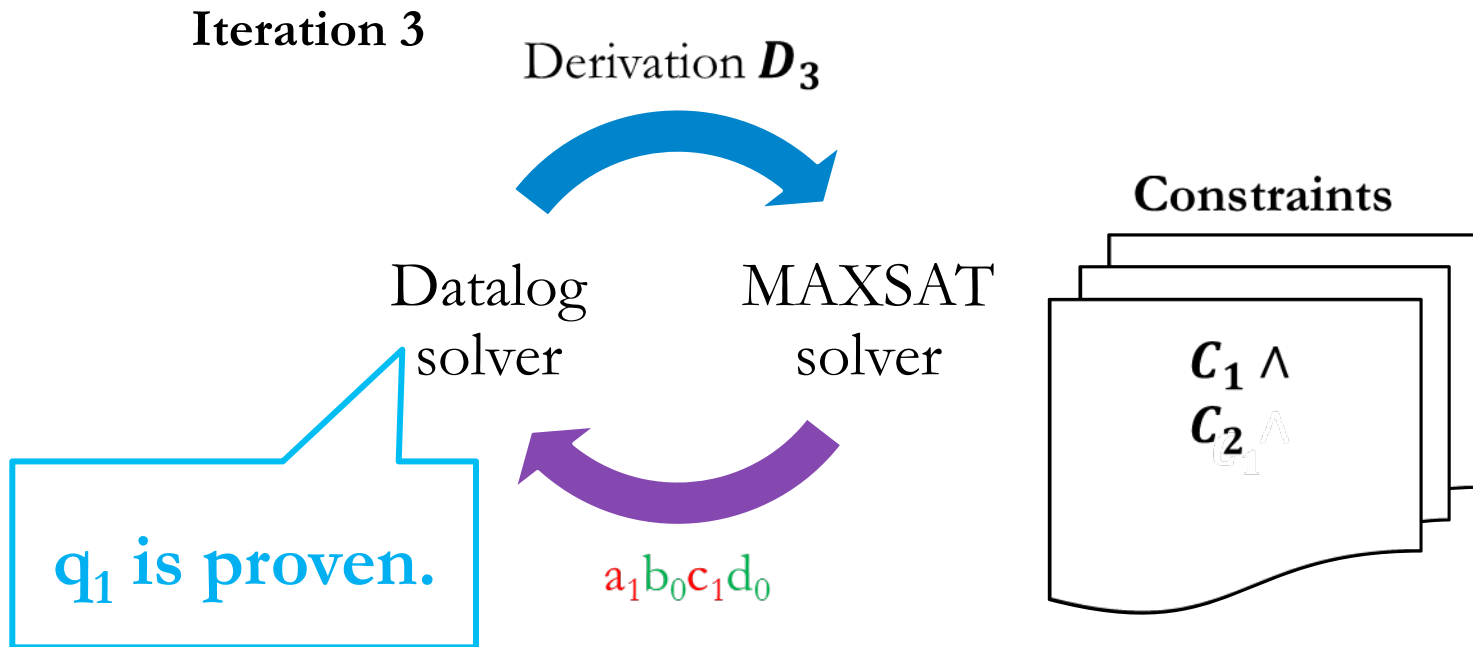
Iteration 2 and Beyond

Iteration 2



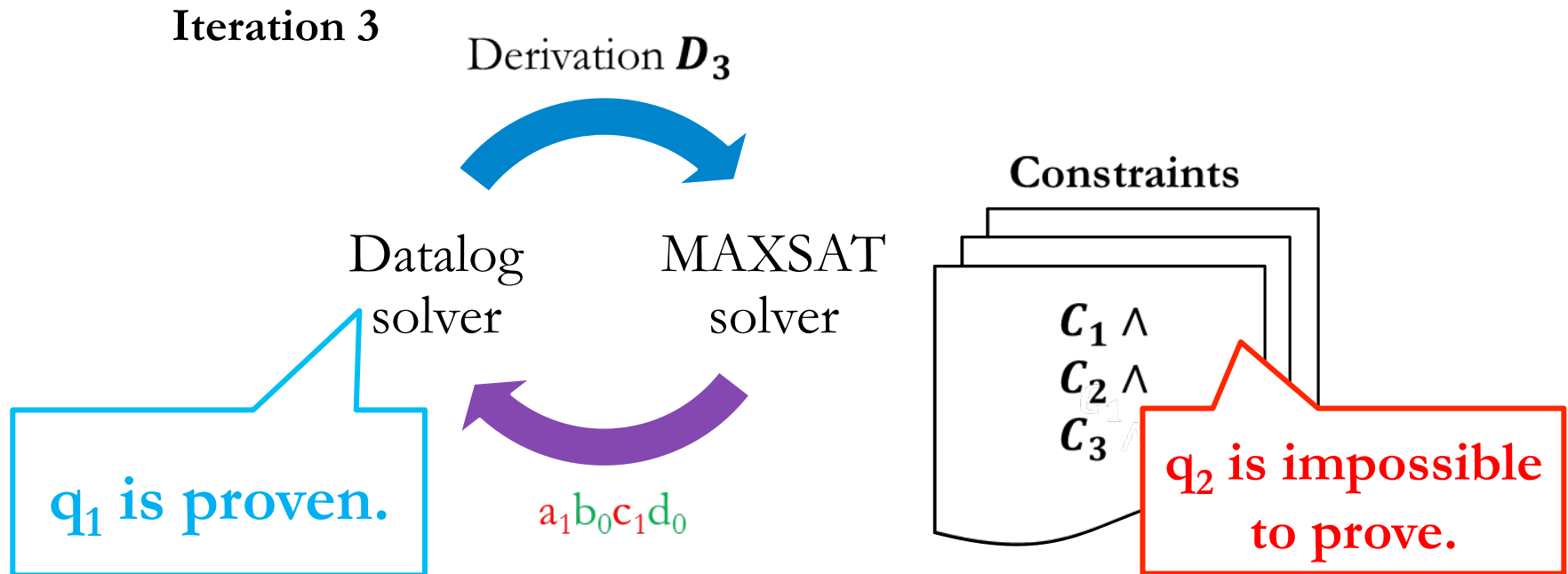
Query	Answer	Eliminated Abstractions
q_1 : path(0, 5)		$a_0c_0d_0, a_0b_0d_0, a_1c_0d_0$ (6/16)
q_2 : path(0, 2)		a_0c_0, a_1c_0 (8/16)

Iteration 2 and Beyond



Query	Answer	Eliminated Abstractions
q_1 : path(0, 5)	✓ $a_1 b_0 c_1 d_0$	$a_0 c_0 d_0, a_0 b_0 d_0, a_1 c_0 d_0$ (6/16)
q_2 : path(0, 2)		$a_0 c_0, a_1 c_0$ (8/16)

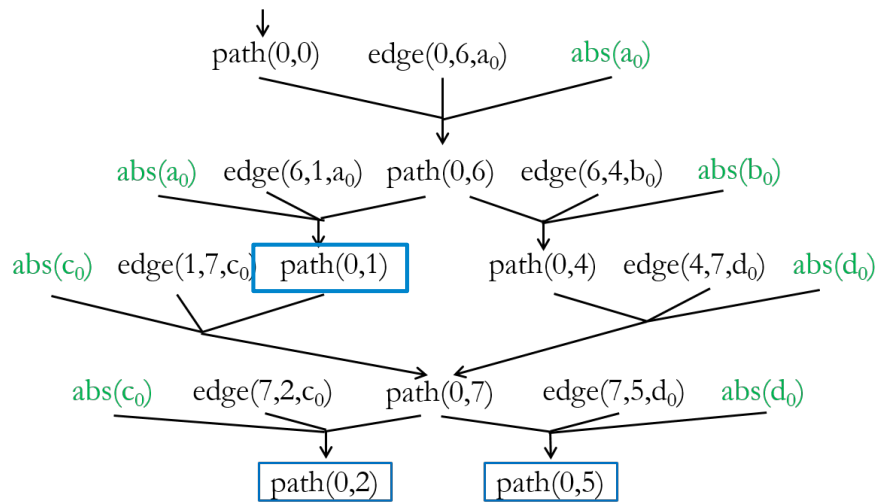
Iteration 2 and Beyond



Query	Answer	Eliminated Abstractions
q_1 : path(0, 5)	✓ $a_1 b_0 c_1 d_0$	$a_0 c_0 d_0, a_0 b_0 d_0, a_1 c_0 d_0$ (6/16)
q_2 : path(0, 2)	✗ Impossibility	$a_0 c_0, a_1 c_0, a_1 c_1, a_0 c_1$ (16/16)

Mixing Counterexamples

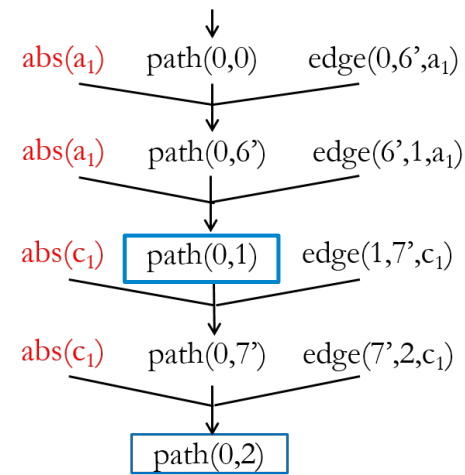
Iteration 1



Eliminated
Abstractions:

$a_0 * c_0 *$

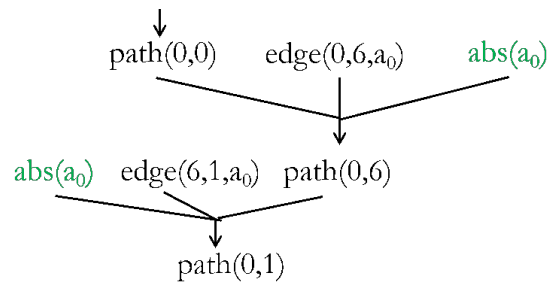
Iteration 3



$a_1 * c_1 *$

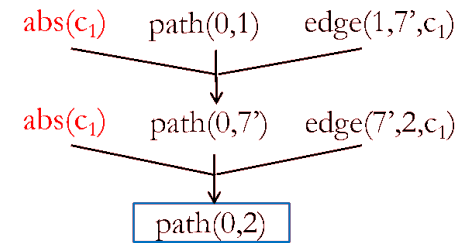
Mixing Counterexamples

Iteration 1



Mixed!

Iteration 3



Eliminated
Abstractions:

$a_0 * c_0 *$

$a_0 * c_1 *$

$a_1 * c_1 *$

Experimental Setup

- ▶ Implemented in JChord using off-the-shelf solvers:
 - ▶ Datalog: bddbddb
 - ▶ MAXSAT: MiFuMaX
- ▶ Applied to two analyses that are challenging to scale:
 - ▶ k-object-sensitivity pointer analysis:
 - ▶ flow-insensitive, weak updates, cloning-based
 - ▶ typestate analysis:
 - ▶ flow-sensitive, strong updates, summary-based
- ▶ Evaluated on 8 Java programs from DaCapo and Ashes.

Benchmark Characteristics

	classes	methods	bytecode(KB)	KLOC
toba-s	1K	6K	423	258
javasrc-p	1K	6.5K	434	265
weblech	1.2K	8K	504	326
hedc	1K	7K	442	283
antlr	1.1K	7.7K	532	303
luindex	1.3K	7.9K	508	295
lusearch	1.2K	8K	511	314
schroeder-m	1.9k	12K	708	460

Results: Pointer Analysis

	queries					iterations
	total	resolved		size		
		current	baseline	final	max	
toba-s	7			170	18K	10
javasrc-p	46			470	18K	13
weblech	5	5	2	140	31K	10
hedc	47	47	6	730	29K	18
antlr	143	143	5	970	29K	15
luindex	138	138	67	1K	40K	26
lusearch	322	322	29	1K	39K	17
schroeder-m	51	51	25	450	58K	15

4-object-sensitivity
< 50%

< 3% of max

Performance of Datalog: Pointer Analysis

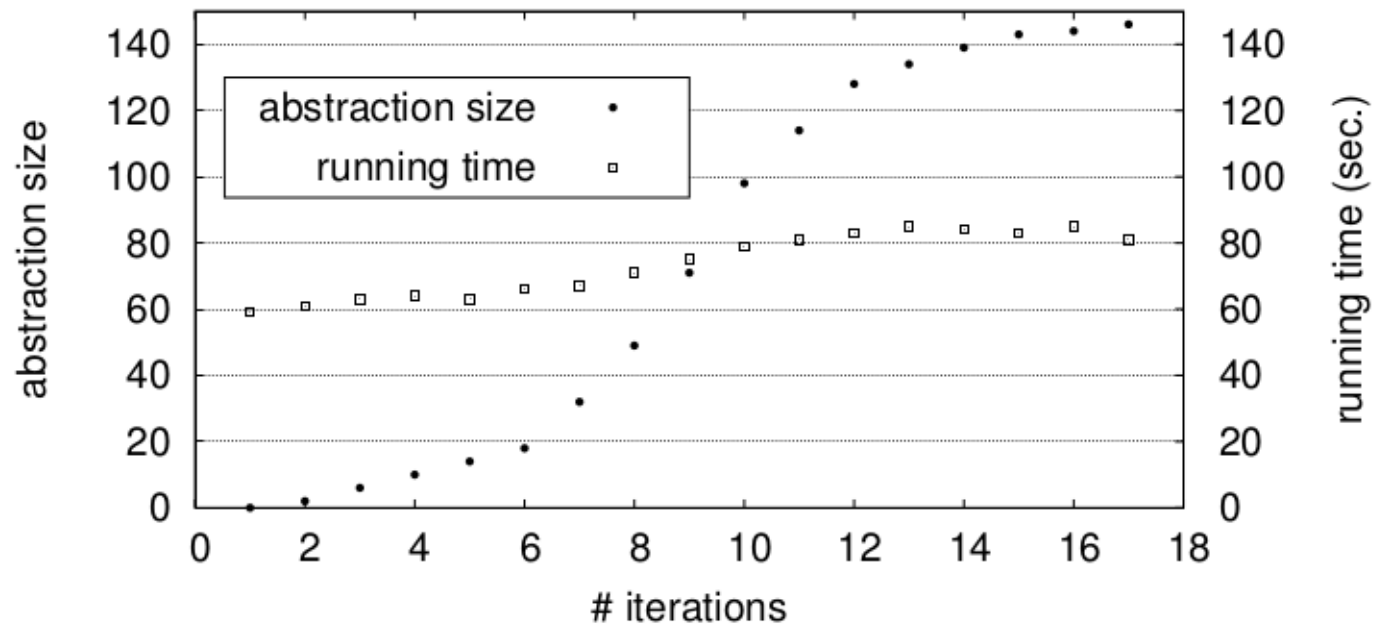
k = 4, 3h28m

Baseline k = 3, 590s

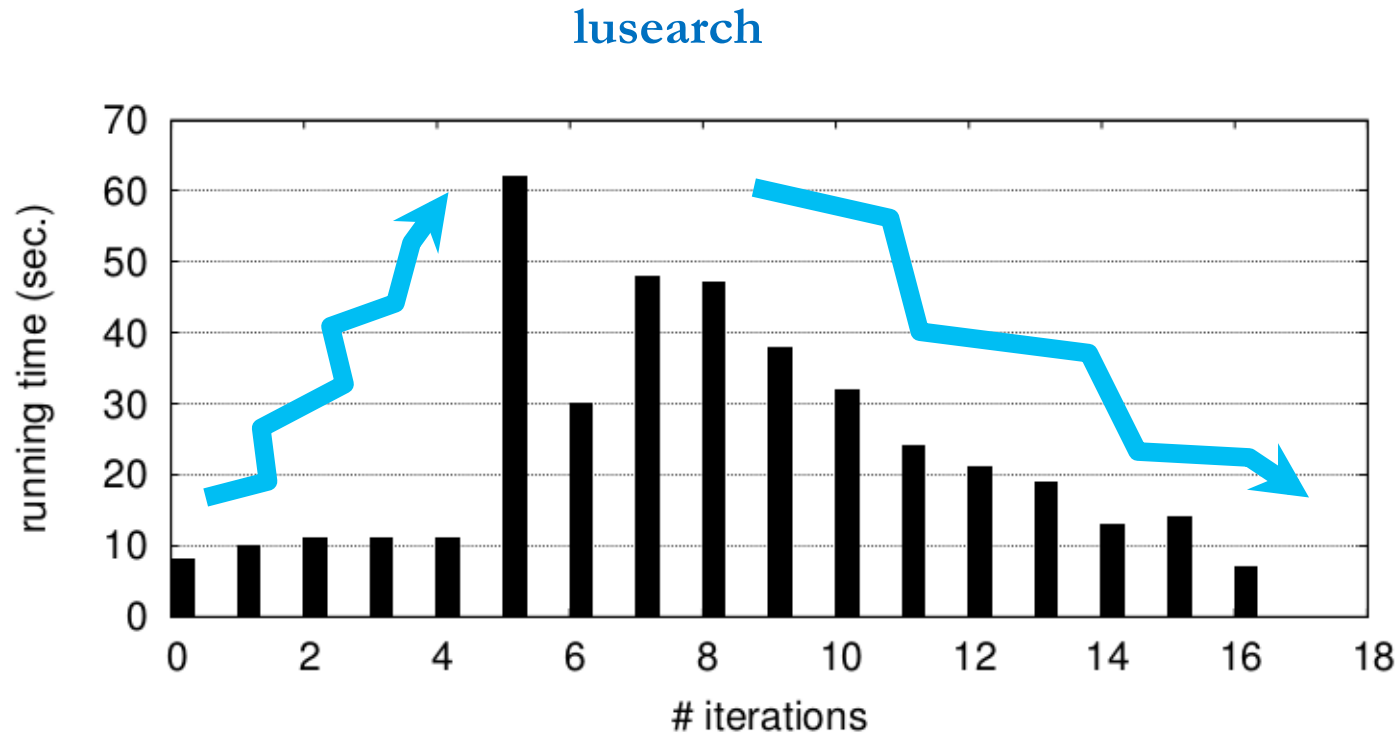
k = 2, 214s

lusearch

k = 1, 153s



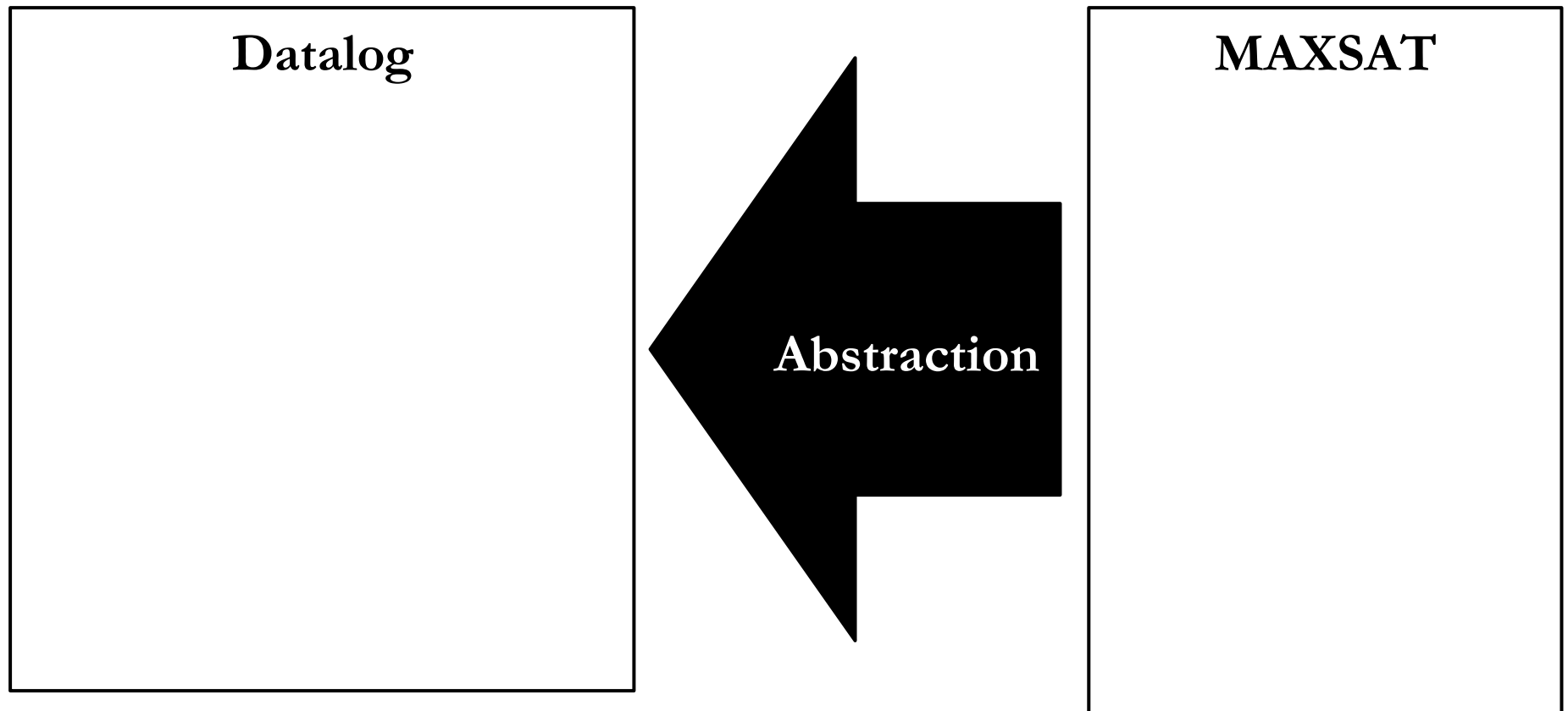
Performance of MAXSAT: Pointer Analysis



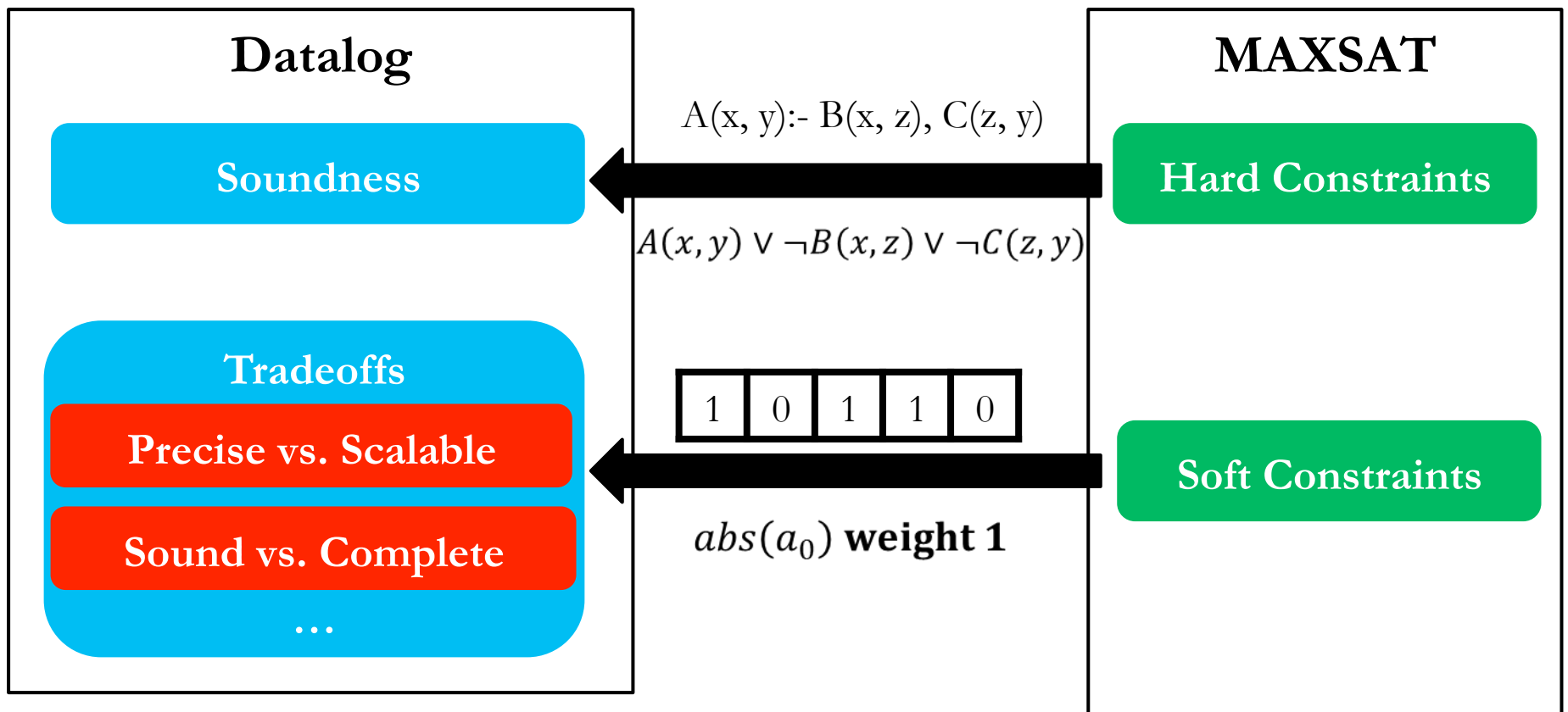
Statistics of MAXSAT Formulae

	pointer analysis	
	variables	clauses
toba-s	0.7M	1.5M
javasrc-p	0.5M	0.9M
weblech	1.6M	3.3M
hedc	1.2M	2.7M
antlr	3.6M	6.9M
luindex	2.4M	5.6M
lusearch	2.1M	5M
schroeder-m	6.7M	23.7M

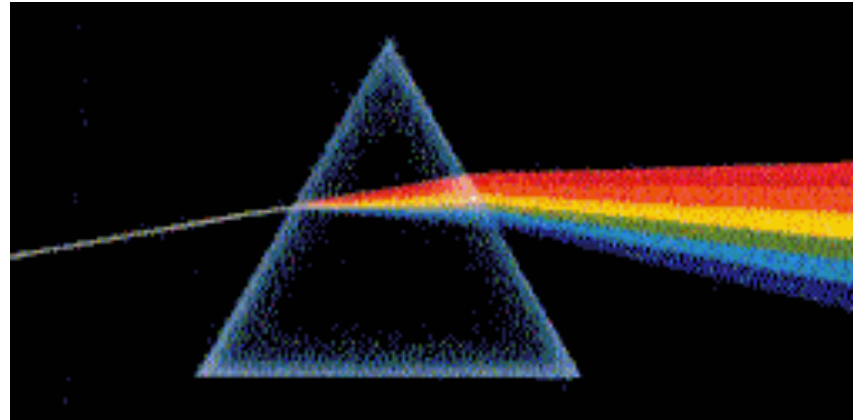
Conclusion



Conclusion



Thank You!



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