

Challenges in State-of-the-Art Bit-Precise Reasoning

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Satisfiability Modulo Theories (SMT)

- » Deciding the **satisfiability of FOL** formula with respect to background **theories**
 - ▶ fixed interpretation of **theory symbols**
 - ▶ Boolean + **domain-specific** reasoning
 - ▶ more expressive than SAT

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- ▶ **Focus: Theory of fixed-size bit-vectors**

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- ▶ **Focus: Theory of fixed-size bit-vectors**

Example. $(x \ll 001) \geq_s 000 \wedge x <_u 100 \wedge (x \cdot 010) \bmod 011 = x + 001$

satisfiable with $x := 001$

Fixed-Size Bit-Vectors

- ▶ **Theory of Fixed-Size Bit-Vectors** provides **bit-precise semantics**
- ▶ **Finite** domain
- ▶ Semantics of arithmetic operations **different** from Real/Int (**modulo 2^n**)

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Example. 8-bit integers (`int8_t`), **counterexample:** $a = 127$, $b = -128$

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» **overflow:** $-(-128) = -128$, thus $a - b = -1$

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Theory of Fixed-Size Bit-Vectors

► **bit-vector** as a sequence of bits of a fixed length

○ **constants, variables:** 010 , $2_{[3]}$, $x_{[3]}$

○ **bit-vector** operators:

● **predicates:** $\langle _ \rangle_u$, $\langle _ \rangle_s$, $\dots$

● **bit-wise:** $\sim$, $\&$, $|$, $\oplus$, $\dots$

● **shift:** $\ll$, $\gg$, $\gg_a$, rotate

● **word:** $\circ$, $[:]$, $\langle _ \rangle_u$, $\langle _ \rangle_s$, repeat

● **arithmetic:** $-$, $+$, $\cdot$, $\div$, mod , $\dots$

● **overflow predicates** for arithmetic operators

► **arithmetic** operators modulo 2^n (**overflow semantics!**)

► **natural representation** for **machine integers**, **hardware registers**

► widely used in **hardware and software verification**

Theory of Fixed-Size Bit-Vectors: Semantics

Most vs. Least Significant Bit:

LSB: $x_3x_2x_1x_0$

MSB: $x_3x_2x_1x_0$

Value Ranges for Size n :

- unsigned: $[0, 2^n - 1]$ $[0, 15]$
- signed: $[-2^{n-1}, 2^{n-1} - 1]$ $[-8, 7]$

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Unsigned vs. Signed Interpretation:

$$1100_u = 8 + 4 = 12$$

$$1100_s = -8 + 4 = -4$$

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$$0011_s = 2 + 1 = 3$$

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Arithmetic

$$8_{[4]} + 11_{[4]} = 3_{[4]}$$

$$8_{[4]} \cdot 2_{[4]} = 0_{[4]}$$

$$-8_{[4]} \div_s -1_{[4]} = -8_{[4]}$$

Division by Zero

$$a_{[n]} \div 0_{[n]} = \sim 0_{[n]} = 1111 \dots 1111$$

$$a_{[n]} \bmod 0_{[n]} = a_{[n]}$$

» division is total function

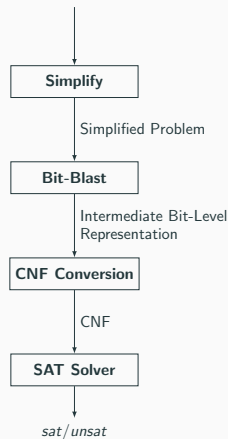
► **The State of The Art** for **quantifier-free** BV formulas

» rewriting + simplifications + **reduction** to SAT

» BV terms » intermediate representation » CNF

► **efficient** in practice (**state-of-the-art SAT solvers**)

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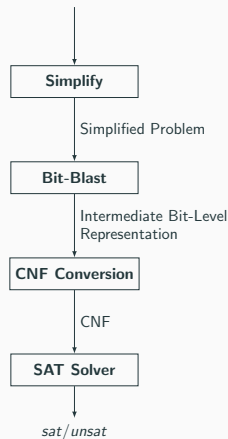
» BV terms » intermediate representation » CNF

► **Main** technique in **state-of-the-art SMT solvers**

◦ *Bitwuzla, Boolector, cvc5, MathSAT5, STP, Yices2, Z3, ...*

► **efficient** in practice (**state-of-the-art SAT solvers**)

$$(x \ll 001) \geq_s 000 \wedge x <_u 100 \\ \wedge (x \cdot 010) \bmod 011 = x + 001$$



Bit-Blasting: Simplify

1. Simplify Formula

► Term rewriting (**local** simplifications)

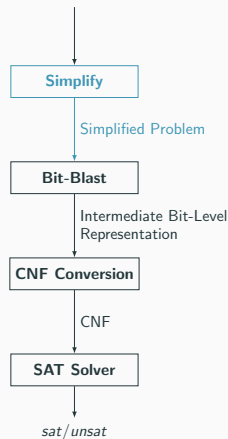
- $a_{[n]} + 0_{[n]} \rightsquigarrow a$
- $a_{[n]} \cdot 0_{[n]} \rightsquigarrow 0$
- ...

» SMT solvers implement **hundreds** of rewriting rules

► Preprocessing (**global** simplifications)

- $a = t \wedge F[a] \rightsquigarrow F[t]$
- ...

$$(x \ll 001) \geq_s 000 \wedge x <_u 100 \\ \wedge (x \cdot 010) \bmod 011 = x + 001$$

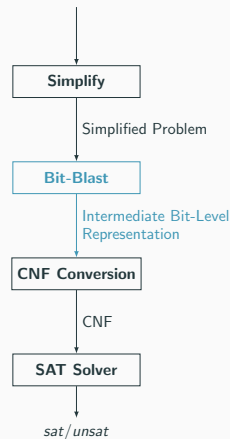


Bit-Blasting: Intermediate Representation

2. Translation into **Intermediate** Bit-Level Representation

- ▶ preserve **hierarchical** bit-level structure
 - » **simplify** bit-level structure **prior** to flattening into CNF
- ▶ bit-level representation can get **very large**
 - » **efficient** and **compact** representation necessary
 - **And-Inverter Graphs (AIG)** » Bitwuzla, Boolector, MathSAT5, STP
 - Xor/Or/Not Graphs » Yices2
 - Arbitrary Boolean » cvc5, Z3

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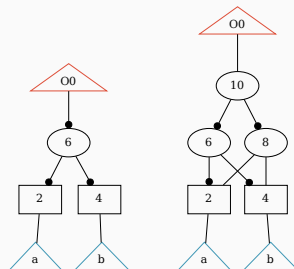


Bit-Blasting: Intermediate Representation

2. Translation into **Intermediate** Bit-Level Representation

► **And-Inverter Graphs (AIG)**

- » directed acyclic graphs over $\wedge, \neg$
 - nodes represent **AND gates**
 - **negation** as edge attribute
- structural hashing: detect and share identical subgraphs
- compact and scalable representation
- efficient simplifications [Mishchenko'06, Brummayer'06]

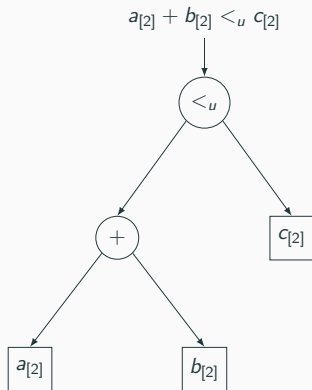


$$a \vee b \\ \equiv \neg(\neg a \wedge \neg b)$$

$$a \oplus b \\ \equiv \neg(\neg a \wedge \neg b) \\ \wedge \neg(a \wedge b)$$

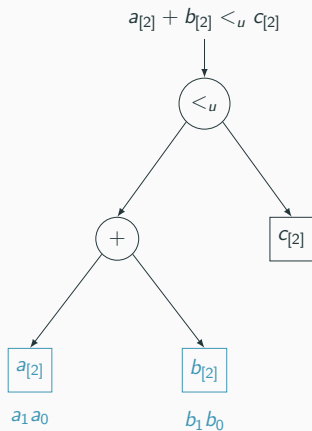
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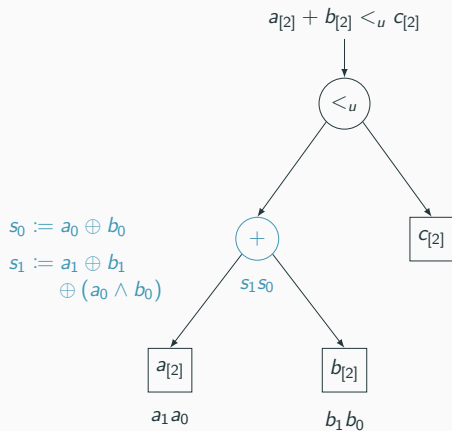
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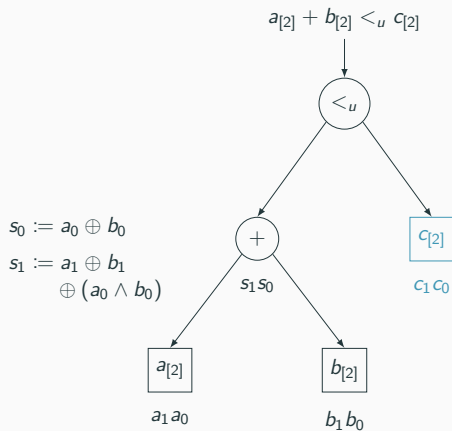
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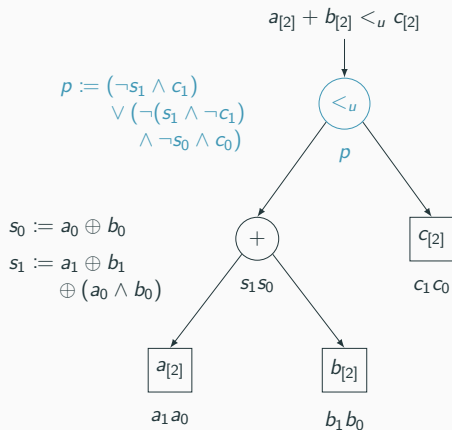
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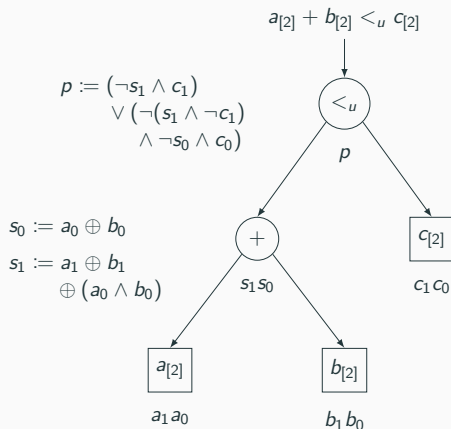
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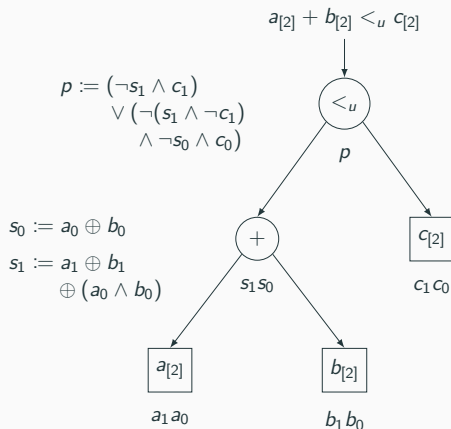
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$$\begin{aligned} p & \\ \wedge p & \leftrightarrow ((\neg s_1 \vee c_1) \\ & \vee (\neg(s_1 \wedge \neg c_1) \\ & \wedge \neg s_0 \wedge c_0)) \\ \wedge s_0 & \leftrightarrow (a_0 \oplus b_0) \\ \wedge s_1 & \leftrightarrow (a_1 \oplus b_1 \\ & \oplus (a_0 \wedge b_0)) \end{aligned}$$

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» translate to CNF via **Tseitin** transformation

3. CNF Conversion via Tseitin Transformation

- introduce fresh variable for each AND-gate:

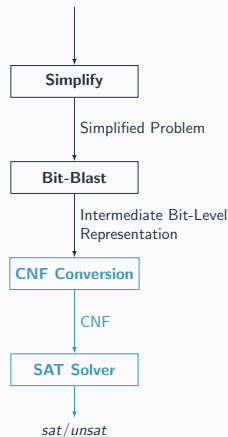
$$\gg x \leftrightarrow a \wedge b$$

- transform into CNF:

$$\gg x \leftrightarrow a \wedge b \rightsquigarrow (\neg x \vee a) \wedge (\neg x \vee b) \wedge (x \vee \neg a \vee \neg b)$$

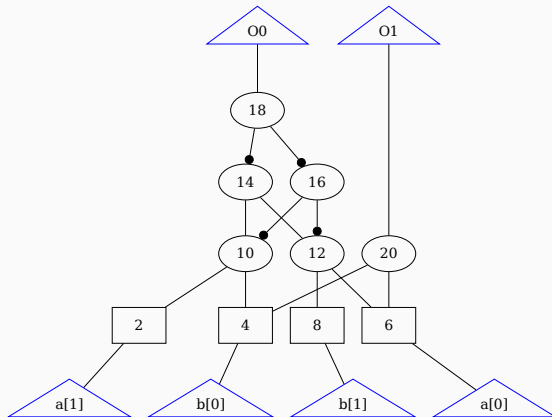
4. Send CNF to SAT Solver

$$(x \lll 001) \geq_s 000 \wedge x <_u 100 \\ \wedge (x \cdot 010) \bmod 011 = x + 001$$



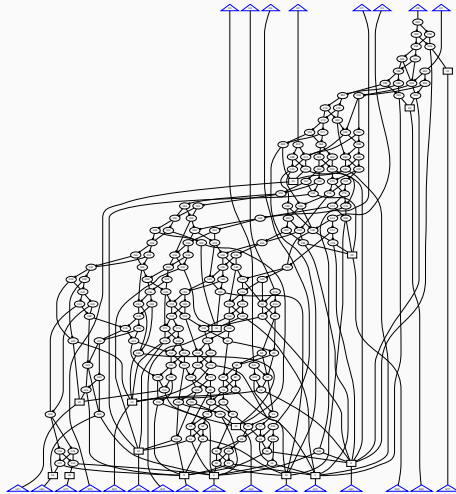
Scalability of Bit-Blasting

2-bit Multiplier (AIG circuit)



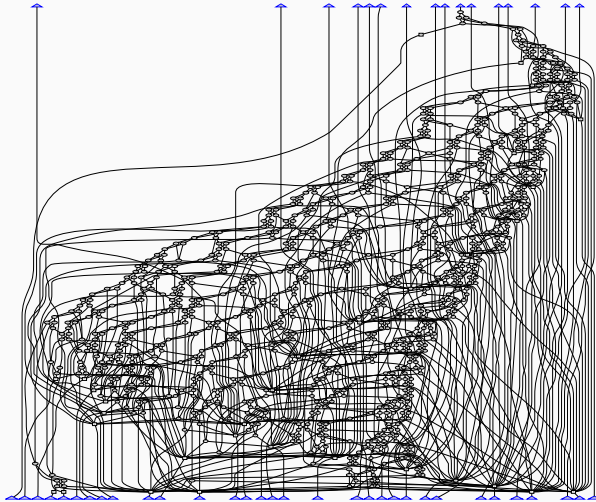
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8-bit Multiplier (AIG circuit)



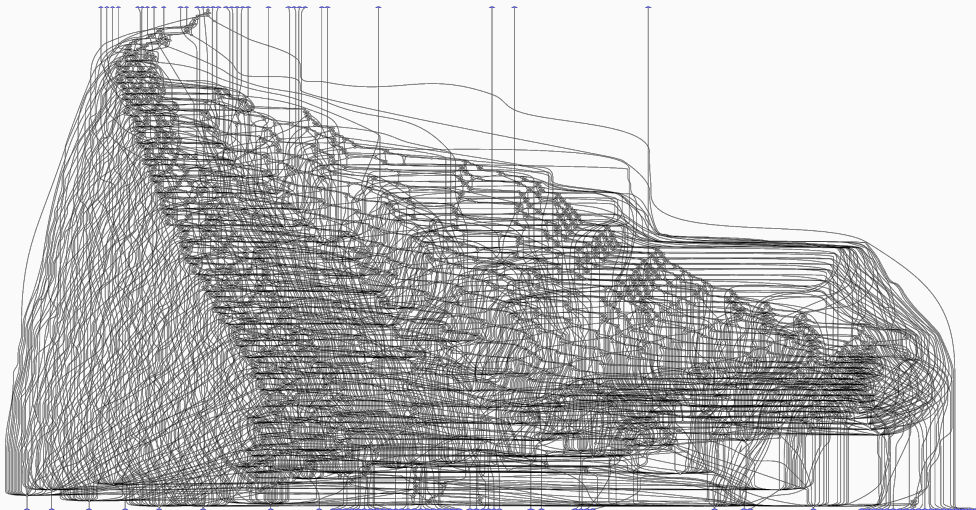
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16-bit Multiplier (AIG circuit)



Scalability of Bit-Blasting

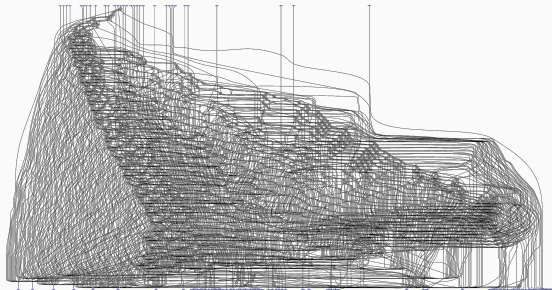
32-bit Multiplier (AIG circuit)



Scalability of Bit-Blasting

- ▶ **does not generally scale well** with **increasing** bit-width
- ▶ especially with **arithmetic** operators
 - » large and complex Boolean circuits
- ▶ potential **bottleneck** for SAT solver
 - » in practice already as low as 32 bits
- ▶ especially severe for applications with **large bit-vectors**
 - **smart contract verification**: 256 bits, heavy use of $\{., \div, \text{mod}\}$
 - **floating-point arithmetic** when word-blasting: 106 bit bvmul for Float64 fp.mul
 - **cryptography**

32-bit Multiplier (AIG circuit)



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 - ▶ **PolySAT** [Rath et al. 2024]
 - CDCL(T) solver in Z3 for **non-linear bit-vector polynomials**

Every Technique Has Scalability Issues

1,500 benchmarks* instantiated with bit-widths 16,...,8192, ~ 85 sat, ~ 1415 unsat

bw	Solved Benchmarks					VBS
	Bit-Blast Bitwuzla	Lazy+Layered CVC4	MCSAT Yices2	Int-Blast cvc5	PolySAT Z3	
16	1,495					
32	1,459					
64	1,440					
128	1,433					
256	1,388					
512	1,277					
1,024	1,065					
2,048	844					
4,096	816					
8,192	744					
99% $\rightarrow$ 49%						

Limits: 1,200 seconds, 8GB memory

* syrew benchmarks from [Niemetz et al. 2024]. 500 term and formula equivalence checks enumerated with cvc5's SyGuS solver using SyGuS grammar $\{0, 1, x, s, t, \approx, \not\approx, <_u, \leq_u, \sim, \&, \ll, \gg, \diamond\}$ for $\diamond \in \{\cdot, \div, \text{mod}\}$.

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16	1,495	1,458	1,394	1,116	696	
32	1,459	1,390	1,194	1,102	672	
64	1,440	1,368	1,112	1,077	668	
128	1,433	1,308	1,076	1,017	648	
256	1,388	1,232	987	916	637	
512	1,277	1,162	916	788	620	
1,024	1,065	774	794	613	608	
2,048	844	401	668	528	576	
4,096	816	300	572	428	562	
8,192	744	202	492	389	552	
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256	1,388	1,232	987	916	637	1,480
512	1,277	1,162	916	788	620	1,421
1,024	1,065	774	794	613	608	1,323
2,048	844	401	668	528	576	1,133
4,096	816	300	572	428	562	1,074
8,192	744	202	492	389	552	993
	99% $\rightarrow$ 49%	97% $\rightarrow$ 13%	93% $\rightarrow$ 33%	74% $\rightarrow$ 26%	46% $\rightarrow$ 37%	100% $\rightarrow$ 66%

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How to tackle these scalability issues?

► Improve scalability of bit-blasting itself

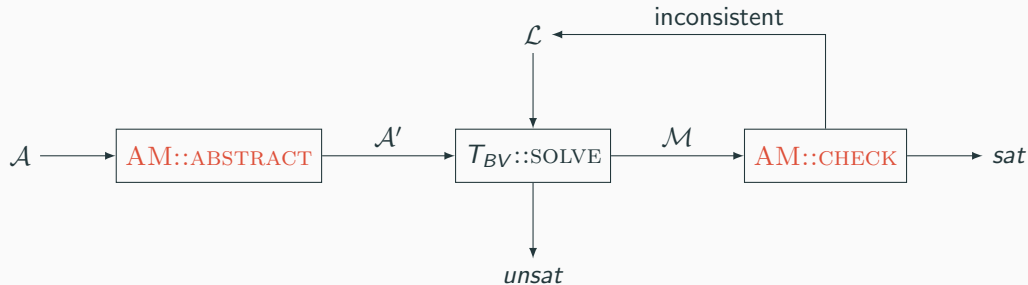
- combination of **under- and over-approximation** [Bryant et al. 2007]
 - **under-approximation** via restricting value ranges of inputs
 - **over-approximation** via ITE elimination, assertion abstraction and abstracting $x \cdot y$ with partially interpreted function $\lambda x. \lambda y. \text{ite}(x \approx 0 \vee y \approx 0, 0, \text{ite}(x = 1, y, \text{ite}(y \approx 1, x, f(x, y))))$
- similar **under-approximation** techniques in [Brummayer 2009, Jonás et al. 2020]
- **CEGAR-style framework** abstracting $\{., \div, \text{mod}\}$ [Niemetz et al. 2024]

Scalable Bit-Blasting with Abstractions*

- ▶ a **CEGAR-style** abstraction-refinement framework for **theory BV**
 - » Counter Example Guided Abstraction Refinement (CEGAR)
 - » based on **bit-blasting**
 - implemented in our SMT solver **Bitwuzla**
 - significantly **improves** performance on **large** bit-vectors
- ▶ **abstraction refinement scheme** for bit-vector operators $\{\cdot, \div, \text{mod}\}$
 - » **most expensive** operators for bit-blasting
 - 70 lemmas total
- ▶ abduction-based framework for **lemma synthesis**
- ▶ lemma **scoring** scheme

* [Niemetz et al. 2024] joint work with M. Preiner and Y. Zohar

Abstraction-Refinement Loop



- abstract each relevant $\{\cdot, \div, \text{mod}\}$ term with **fresh constant**
- **over-approximation**

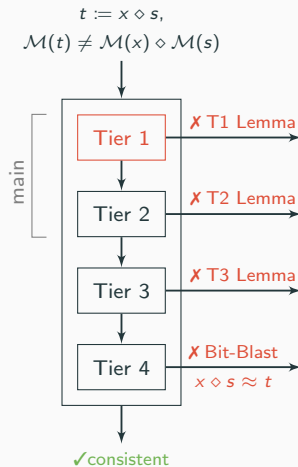
- **check consistency** with true semantics of operators
 - » **consistent**: ✓
 - » **inconsistent**: refine abstraction

4-Tiered Refinement Scheme

► CEGAR Loop Refinement Step for Abstracted Term

For each abstracted term $t := x \diamond s$

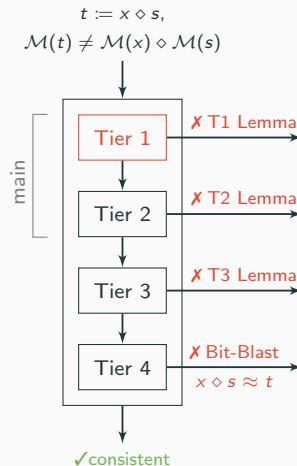
1. check if $\mathcal{M}(t) \neq \mathcal{M}(x) \diamond \mathcal{M}(s)$
 - » e.g., $t := x \cdot s$ $\mathcal{M} = \{x_{[32]} = 3, s_{[32]} = 6, t_{[32]} = 1\}$
2. check if a lemma in tiers 1–3 evaluates to false under $\mathcal{M}$
 - » tiers processed in order 1-4
3. use first violated lemma as refinement



4-Tiered Refinement Scheme

► Tier 1 Hand-Crafted Lemmas

- ▷ properties described by **invertibility condition** IC
- Given $x \diamond s \approx t$, is there a value for x given s and t ?
 - $\exists x. x \diamond s \approx t \Leftrightarrow IC[s, t]$

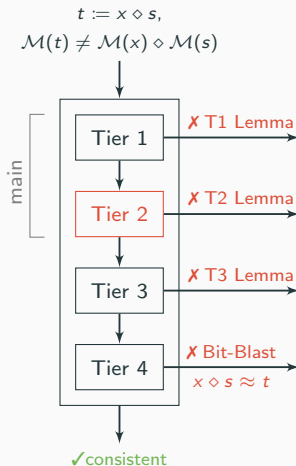
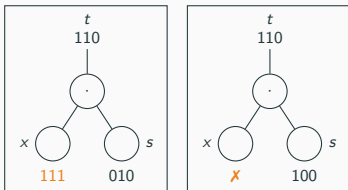


4-Tiered Refinement Scheme

► Tier 1 Hand-Crafted Lemmas

▷ properties described by **invertibility condition IC**

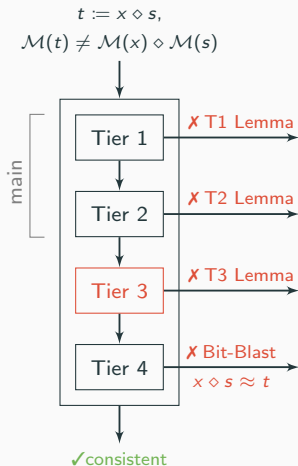
- Given $x \diamond s \approx t$, is there a value for x given s and t ?
- $\exists x. x \diamond s \approx t \Leftrightarrow IC[s, t]$
- e.g., $x \cdot s$ has at least as many trailing zeros as x or s
 - » $((-s \mid s) \& t) \approx t$
 - » $((-x \mid x) \& t) \approx t$



4-Tiered Refinement Scheme

► Tier 1 Hand-Crafted Lemmas

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 - e.g., $x \cdot s$ has at least as many trailing zeros as x or s
 - » $((-s \mid s) \& t) \approx t$
 - » $((-x \mid x) \& t) \approx t$
- ▷ **basic** properties of the abstracted operator
 - e.g., abstraction t for $x \cdot s$ is odd iff x and s are odd
 - » $t[0] \approx (x[0] \& s[0])$
- ▷ **verified** up to bit-width 512 (17 lemmas)



4-Tiered Refinement Scheme

► Tier 2 Synthesized Lemmas

▷ synthesized via **syntax-restricted abduction** with **cvc5** (offline)

» **abduction**: A, B , find C s.t. $A \wedge C \Rightarrow B$ valid, $A \wedge C$ sat

» find **lemma** ℓ such that $\top \wedge \neg \ell \Rightarrow x \diamond s \not\approx t$

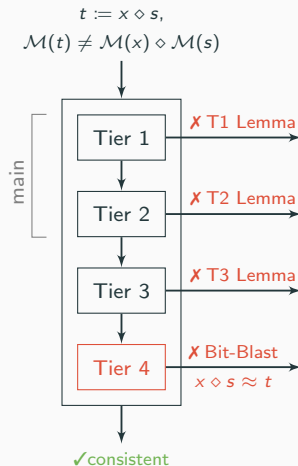
▷ e.g., for $x \cdot s \approx t$

» $x \not\approx (1 \oplus (x \ll (s \oplus t)))$

▷ lemmas **filtered** based on **lemma score**

▷ **with respect to hand-crafted (tier 1) lemmas**

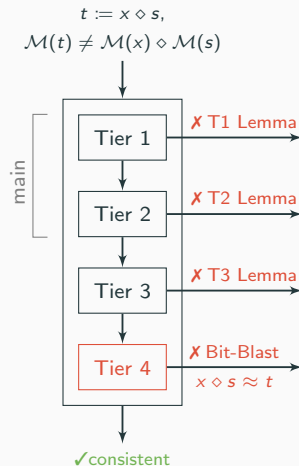
▷ **verified** up to bit-width 512 (53 lemmas)



4-Tiered Refinement Scheme

► Tier 3 Value Instantiation Lemmas

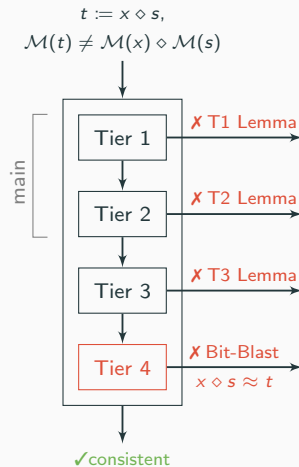
- ▷ rule out **current inconsistent model value**
- ▷ only added if none in tiers 1–2 are violated
- ▷ e.g., for $x \cdot s$ and $\mathcal{M} = \{x_{[32]} = 3, s_{[32]} = 6, t_{[32]} = 1\}$,
we add lemma $(x = 3 \wedge s = 6) \Rightarrow t = 18$
- ▷ **limited** fallback strategy
 - » limited to $w/8$ instantiations per term
 - » e.g., 4 instantiations for 32 bit



4-Tiered Refinement Scheme

► Tier 4 Bit-blasting lemmas

- ▷ last resort
- ▷ add lemma to **enforce bit-blasting** of the abstracted term
- ▷ e.g., lemma $t \approx x \cdot s$ for $x \cdot s$



Tier 1* and 2 Refinement Lemmas ($t := x \diamond s$)

bvmul

1* $s \approx 2^i \Rightarrow t \approx x \ll i$	11 _{>1} $t \not\approx (1 \mid \sim(x \oplus s))$
2* $s \approx -2^i \Rightarrow t \approx -x \ll i$	12 _{>1} $t \not\approx (\sim 1 \mid (x \oplus s))$
3* $((-s \mid s) \& t) \approx t$	13 $x \not\approx ((x \ll (s+t)) - 1)$
4* $t[0] \approx (x[0] \& s[0])$	14 $x \not\approx (1 - (x \ll (s-t)))$
5 _{>1} $s \not\approx \sim(t \mid (1 \& (x \mid s)))$	15 $s \not\approx (1 + (s \ll (t-x)))$
6 _{>1} $(x \& t) \not\approx (s \mid \sim t)$	16 $s \not\approx (1 - (s \ll (t-x)))$
7 _{>1} $t \not\approx ((s \mid 1) \ll (t \ll x))$	17 $s \not\approx (1 + (s \ll (x-t)))$
8 $s \approx (s \ll (x \& (1 \gg t)))$	18 _{>1} $t \not\approx (1 \mid (x+s))$
9 _{≠2} $t \geq_u (1 \& ((x \& s) \gg 1))$	19 $x \not\approx \sim(x \ll (s+t))$
10 $x \not\approx (1 \oplus (x \ll (s \oplus t)))$	

bvurem

1* $s \approx 2^i \Rightarrow t \approx (0_{[i \ll (x) - i]} \circ x[i-1 : 0])$	9 $x \geq_u (t \mid (x \& s))$
2* $s \not\approx 0 \Rightarrow t \leq_u s$	10 $1 \not\approx (t \& \sim(x \mid s))$
3* $x \approx 0 \Rightarrow t \approx 0$	11 $t \not\approx (\sim x \mid -s)$
4* $s \approx 0 \Rightarrow t \approx x$	12 $(t \& (x \mid s)) \geq_u (t \& 1)$
5* $s \approx x \Rightarrow t \approx 0$	13 _{>2} $x \not\approx (-x \mid \sim t)$
6* $x <_u s \Rightarrow t \approx x$	14 $(x + -s) \geq_u t$
7* $\sim -s \geq_u t$	15 $(-s \oplus (x \mid s)) \geq_u t$
8 $x \approx (x \& (s \mid (t \mid -s)))$	

bvudiv

1* $s \approx 2^i \Rightarrow t \approx x \gg i$	19 $(x \gg t) \not\approx (s \mid t)$
2* $(s \approx x \wedge s \not\approx 0) \Rightarrow t \approx 1$	20 $s \not\approx \sim(s \gg (t \gg 1))$
3* $s \approx 0 \Rightarrow t \approx \sim 0$	21 _{>1} $x \not\approx \sim(x \& (t \ll 1))$
4* $(x \approx 0 \wedge s \not\approx 0) \Rightarrow t \approx 0$	22 $t \geq_u ((x \ll 1) \gg s)$
5* $s \not\approx 0 \Rightarrow t \leq_u x$	23 $x \geq_u (s \ll \sim(x \mid t))$
6* $(s \approx \sim 0 \wedge x \not\approx \sim 0) \Rightarrow t \approx 0$	24 $x \geq_u (t \ll \sim(x \mid s))$
7 $x \geq_u -(-s \& -t)$	25 $x \geq_u (t \oplus (t \gg (s \gg 1)))$
8 $-(s \mid 1) \geq_u t$	26 $x \geq_u (s \oplus (s \gg (t \gg 1)))$
9 $t \not\approx -(s \& \sim x)$	27 $x \geq_u (s \ll \sim(x \oplus t))$
10 $(s \mid t) \not\approx (x \& \sim 1)$	28 $x \geq_u (t \ll \sim(x \oplus s))$
11 $(s \mid 1) \not\approx (x \& \sim t)$	29 $x \not\approx (t + (s \mid (x + s)))$
12 $(x \& -t) \geq_u (s \& t)$	30 _{>2} $x \not\approx (t + (1 + (1 \ll x)))$
13 $s \geq_u (x \gg t)$	31 $s \geq_u ((x+t) \gg t)$
14 $x \geq_u ((s \gg (s \ll t)) \ll 1)$	32 _{>1} $x \not\approx (t + (t + (x \mid s)))$
15 $x \geq_u ((t \ll 1) \gg (t \ll s))$	33 $(s \oplus (x \mid t)) \geq_u (t \oplus 1)$
16 $t \geq_u ((x \gg s) \ll 1)$	34 $t \geq_u (x \gg (s-1))$
17 $x \geq_u ((x \mid t) \& (s \ll 1))$	35 $(s-1) \geq_u (x \gg t)$
18 $x \geq_u ((x \mid s) \& (t \ll 1))$	36 _{≠2} $x \not\approx (1 - (x \ll (x-t)))$

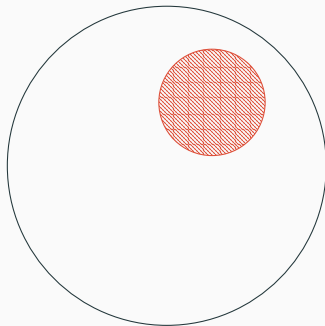
Lemma Score

► Metric for **quality** of a lemma

- ▷ Given abstracted term $x \diamond s$ and lemma $\ell[x, s, t]$ s.t. $x \diamond s \approx t \Rightarrow \ell$, then
 $\text{SCORE}(\ell, w) := \text{num. triplets } (v^x, v^s, v^t) \text{ where } \ell[v^x, v^s, v^t] = \top$.

Example. $x \cdot s$ with $w = 4$

- Worst score: $2^4 \times 2^4 \times 2^4 = 4096$ ($\ell = \top$)
- Best score: $2^4 \times 2^4 = 256$ ($\ell = x \cdot s \approx t$)



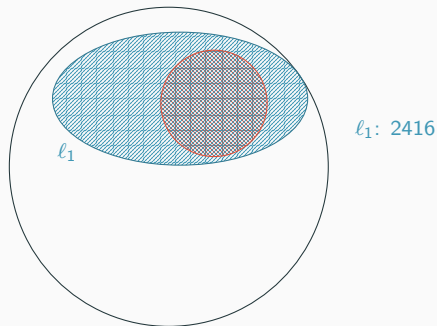
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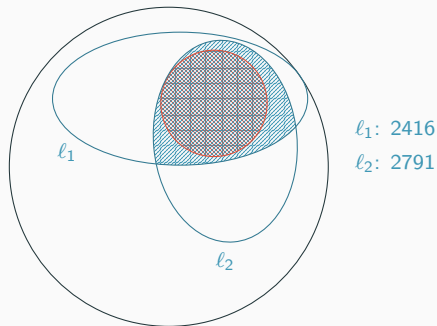
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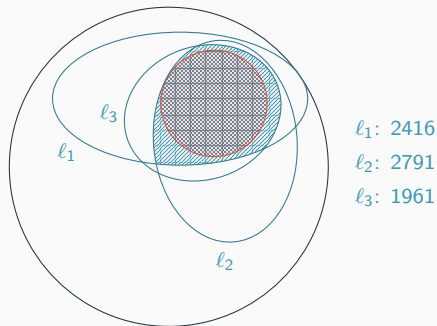
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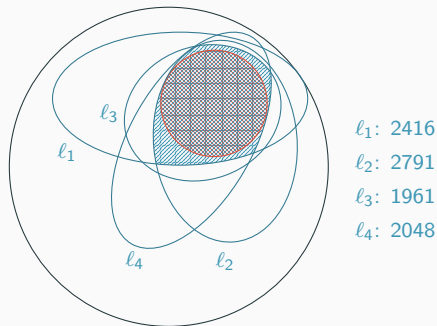
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- **Score for hand-crafted $\{\ell_1, \dots, \ell_4\}$: 704**
 - » rule out 88% of incorrect triplets



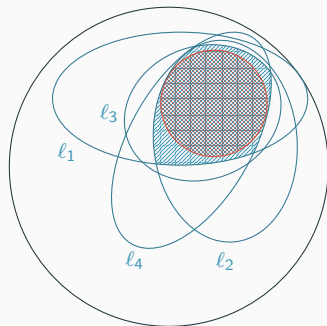
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- **Score for hand-crafted $\{\ell_1, \dots, \ell_4\}$: 704**
 - » rule out **88%** of incorrect triplets
- **Overall scores** (tiers 1+2, $w = 4$)
 - » $x \cdot s$ 490 $\cong$ **94%** (19 lemmas)
 - » $x \div s$ 396 $\cong$ **96%** (36 lemmas)
 - » $x \bmod s$ 400 $\cong$ **96%** (15 lemmas)



ℓ_1 : 2416
 ℓ_2 : 2791
 ℓ_3 : 1961
 ℓ_4 : 2048

Every Technique Has Scalability Issues

1,500 benchmarks* instantiated with bit-widths 16,...,8192, ~ 85 sat, ~ 1415 unsat

bw	Solved Benchmarks						Bit-Blast+Abstr Bitwuzla
	Bit-Blast Bitwuzla	Lazy+Layered CVC4	MCSAT Yices2	Int-Blast cvc5	PolySAT Z3	VBS	
16	1,495	1,458	1,394	1,116	696	1,500	
32	1,459	1,390	1,194	1,102	672	1,490	
64	1,440	1,368	1,112	1,077	668	1,487	
128	1,433	1,308	1,076	1,017	648	1,488	
256	1,388	1,232	987	916	637	1,480	
512	1,277	1,162	916	788	620	1,421	
1,024	1,065	774	794	613	608	1,323	
2,048	844	401	668	528	576	1,133	
4,096	816	300	572	428	562	1,074	
8,192	744	202	492	389	552	993	
	99% $\rightarrow$ 49%	97% $\rightarrow$ 13%	93% $\rightarrow$ 33%	74% $\rightarrow$ 26%	46% $\rightarrow$ 37%	100% $\rightarrow$ 66%	

Limits: 1,200 seconds, 8GB memory

* syrew benchmarks from [Niemetz et al. 2024]. 500 term and formula equivalence checks enumerated with cvc5's SyGuS solver using SyGuS grammar $\{0, 1, x, s, t, \approx, \not\approx, <_u, \leq_u, \sim, \&, \ll, \gg, \diamond\}$ for $\diamond \in \{\cdot, \div, \text{mod}\}$.

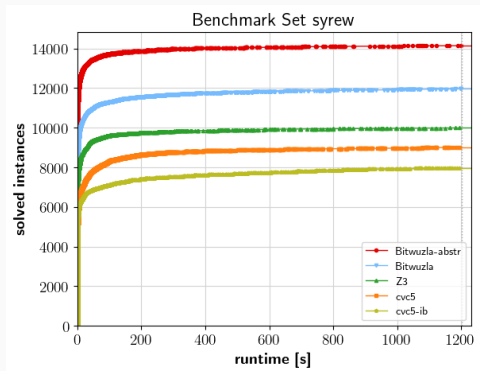
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8,192	744	202	492	389	552	993	1,274
	99% $\rightarrow$ 49%	97% $\rightarrow$ 13%	93% $\rightarrow$ 33%	74% $\rightarrow$ 26%	46% $\rightarrow$ 37%	100% $\rightarrow$ 66%	99% $\rightarrow$ 85%
Limits: 1,200 seconds, 8GB memory						 100% $\rightarrow$ 88%	

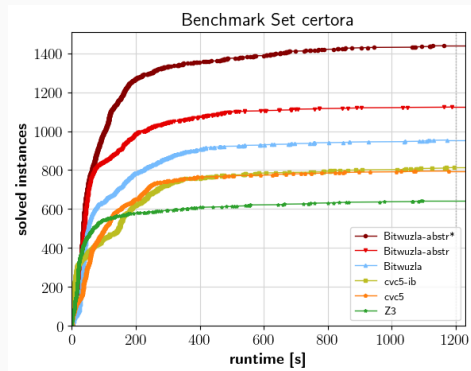
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Evaluation



» 500 term and formula equivalence checks

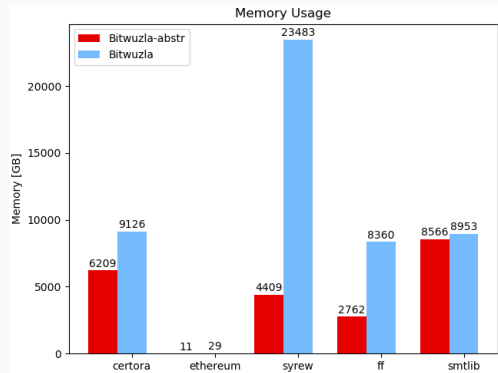
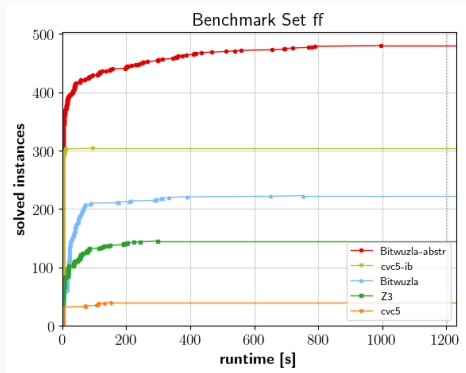
- enumerated by SyGuS (cvc5) for $w = 4$
- instantiated for $w = \{16, 32, \dots, 8192\}$
- 1500 benchmarks, ~ 85 sat, ~ 1415 unsat



» Smart contract verification

- 256 bit bit-vectors
- heavy use of $\{\cdot, \div, \text{mod}\}$

Evaluation



» Translation validation of ZK proofs

- 510 bit bit-vectors

Evaluation: Abstraction Statistics over all Benchmark Sets

► 80% solved without bit-blasting of abstracted terms

» Tier 1-3 refinement lemmas were sufficient to solve the problem

Family	Solved	Abstr. Only	T1	T2	T3	T4
<i>certora₁</i>	255	86.7%	86.3%	78.0%	60.4%	13.3%
<i>certora₂</i>	395	97.2%	85.3%	46.3%	50.4%	2.8%
<i>ethereum</i>	24	100%	41.7%	29.2%	16.7%	0.0%
<i>syrew</i>	7,355	91.9%	98.1%	23.5%	12.3%	8.1%
<i>ff</i>	480	95.2%	84.6%	38.1%	34.8%	4.8%
<i>smtlib</i>	15,524	73.9%	64.3%	27.5%	36.5%	26.1%
Total	24,033	80.4%	75.6%	27.3%	29.5%	19.6%

► 20% required bit-blasting of abstracted terms

» Only 37% of multiplication, 13% of division, 2% of remainder terms bit-blasted

▷ On average 37 refinement iterations (median 4)

What is still challenging?

Example: Commutativity of Multiplication

$$a \cdot b \neq b \cdot a$$

```
(set-logic QF_BV)
(declare-const a (_ BitVec 13))
(declare-const b (_ BitVec 13))
(assert (distinct (bvmul a b) (bvmul b a)))
(check-sat)
```

bw	CNF v/c	Solve [s]	DRAT Size
5	142/282	< 0.00	6.2K
6	216/467	0.03	56K
7	305/697	0.16	593K
8	406/972	0.6	2.2M
9	528/1292	3.4	9.6M
10	662/1657	18	40M
11	811/2067	95	171M
12	975/2522	436	459M
13	1152/3022	2188	1.8G

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13	1152/3022	2188	1.8G

► Arithmetic circuit verification **very hard for SAT solvers**

» results indicate **exponential size resolution proofs** [Biere'16]

► This example: **trivial for SMT** solvers with **commutativity rewrite rule**

What is still challenging?

Example: Multiplication + Bitwise

$$a \cdot b \neq (a \& \sim b) \cdot (\sim a \& b) + (a \& b) \cdot (a \mid b)$$

```
(set-logic QF_BV)
(declare-const a (_ BitVec 16))
(declare-const b (_ BitVec 16))
(assert
  (distinct
    (bvmul a b)
    (bvadd
      (bvmul (bvand a (bvnot b)) (bvand (bvnot a) b))
      (bvmul (bvand a b) (bvor a b)))
  )
)
(check-sat)
```

- ▶ Worse SAT solver behavior as before
- ▶ Very hard for SMT solvers
 - Insufficient word-level reasoning
- ▶ Enumeration faster than SMT solvers
 - For 16-bit example (left)
 - SMT solvers time out after 60min
 - Bit-Blast (Bitwuzla)
 - Int-Blast (cvc5)
 - Lazy Bit-Blast+Layered (CVC4)
 - MCSAT (Yices2)
 - PolySAT (Z3)
 - C++ program enumerates in ~13s

Takeaways

- ▶ **Large bit-vectors** and/or **arithmetic challenging** for state-of-the-art approaches
 - » arithmetic usually **in combination** with bitwise/shift/word operators
 - » scalability **not only** an issue for bit-blasting
- ▶ Bit-Blasting still **best-performing** approach
 - » alternative approaches **complementary**
- ▶ **CEGAR-style** abstraction-refinement for **theory BV** based on **bit-blasting**
 - » significantly improves **scalability** of bit-blasting
 - » for majority of solved instances **full arithmetic circuit not required**

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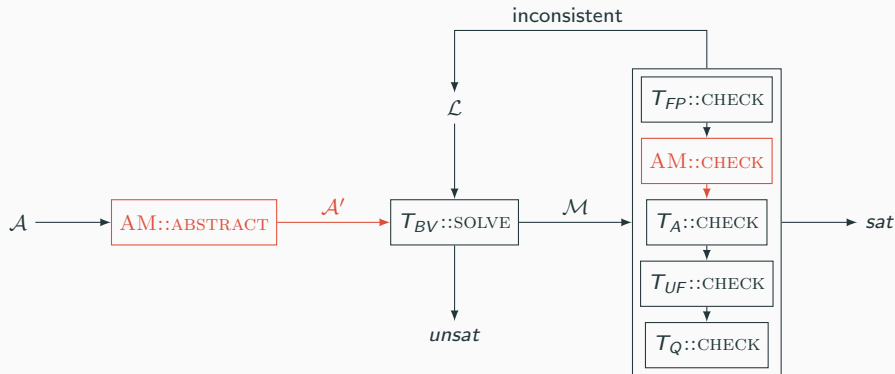
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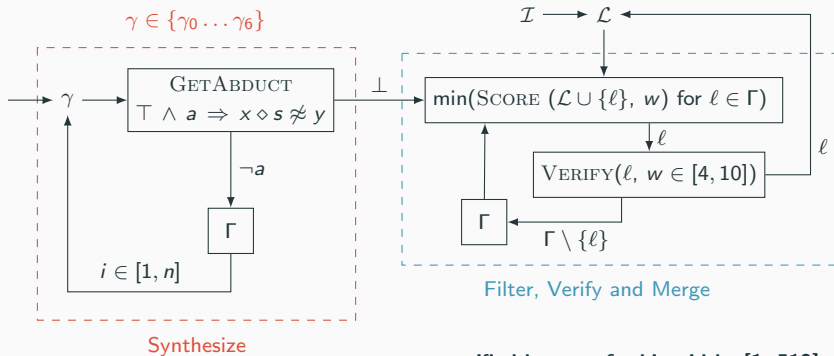
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Bitwuzla: Lemmas on Demand Loop with Abstraction Module



- abstract each $\{\cdot, \div, \text{mod}\}$ term of size ≥ 32 with **fresh constant**
- optional: **assertion abstraction** [?]
 - » interleaved with term abstraction
 - » effective if *unsat* core very small
- order **not arbitrary**
 - » T_{FP} word-blasted to T_{BV}
 - » T_A , T_{UF} and T_Q require consistent T_{BV} abstraction

Lemma Synthesis



► verified lemmas for bit-widths [1, 512]

- Bitwuzla, cvc5, Yices, Z3
- 8 hours time limit, 8 GB memory limit
- 16,896 benchmark, 6348 CPU hours

Evaluation: FP logics

► SMT-LIB

» *smtlib*

- all quantifier-free and quantified logics supported by Bitwuzla (24 in total)
- combinations of BV, FP, UF, Arrays

► FP Logics

» Bitwuzla uses word-blasting for FP logics

- Reduction of FP to bit-vectors with SymFPU [Brain et al. 2019b]

» Operations on unpacked formats require full precision arithmetic

- Float32 fp.mul: 48-bit bvmul
- Float64 fp.mul: 106-bit bvmul

Bitwuzla vs. Bitwuzla-abstr

Logic	Solved (Diff)	Time [s] (Common)
FP	+5	−16%
BVFP	0	−45%
QF_ABVFP	+1	−33%
QF_ABVFPLRA	0	−23%
QF_BVFP	+1	−45%
QF_BVFPLRA	+9	−46%
QF_FP	+23	−13%
QF_FPLRA	+1	−7%
QF_BV/float	+11	−16%

Evaluation

Benchmarks	Solver	Solved	Timeout	Memout	Time [s]	Mem [GB]
<i>certora₁</i> (850)	ABSTR-TA	573	231	46	448k	2,492
	ABSTR-A	386	140	324	681k	5,201
	Bitwuzla-abstr	258	155	437	760k	4,807
	cvc5-ib	147	674	0	879k	667
	Bitwuzla	111	86	653	915k	6,182
	cvc5	90	113	610	923k	6,064
	Z3	30	447	373	989k	4,944
<i>certora₂</i> (1,138)	ABSTR-TA	866	264	8	370k	1,024
	Bitwuzla-abstr	866	263	9	384k	1,402
	ABSTR-A	844	269	25	433k	2,661
	Bitwuzla	843	266	29	439k	2,944
	cvc5	705	223	210	603k	4,027
	cvc5-ib	666	472	0	643k	106
	Z3	612	492	34	679k	1,866
<i>ethereum</i> (3,173)	Bitwuzla-abstr	3,173	0	0	407	11
	Bitwuzla	3,173	0	0	720	29
	Z3	3,169	4	0	6k	107
	cvc5	3,158	0	1	18k	36
	cvc5-ib	3,141	20	0	39k	21

Evaluation

Benchmarks	Solver	Solved	Timeout	Memout	Time [s]	Mem [GB]
<i>syrew</i> (15,000)	Bitwuzla-abstr	14,142	583	276	1,225k	4,409
	Bitwuzla	11,961	744	2,296	3,955k	23,483
	Z3	9,992	833	4,175	6,198k	39,506
	cvc5	9,003	797	5,200	7,498k	48,421
	cvc5-ib	7,974	5,137	1,632	8,836k	19,850
<i>ff</i> (1,224)	cvc5-ff	973	129	122	313k	1,364
	Bitwuzla-abstr	480	729	15	913k	2,762
	cvc5-ib	304	822	98	1,104k	1,074
	Bitwuzla	223	71	930	1,211k	8,360
	Z3	145	56	1,023	1,299k	8,893
	cvc5	40	0	1,184	1,422k	9,523
<i>smtlib</i> (155,269)	Bitwuzla-abstr	148,554	1,944	152	8,770k	8,566
	Bitwuzla	148,492	1,966	193	8,748k	8,953
	Z3	145,121	4,846	565	13,528k	18,278
	cvc5	144,829	3,775	285	13,513k	11,029
	cvc5-ib	127,144	24,479	194	39,647k	15,233

36 unique

Limits: 1200 seconds CPU time limit, 8GB memory limit