

The Power of Extended Resolution

A Practitioner's Perspective

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**Carnegie
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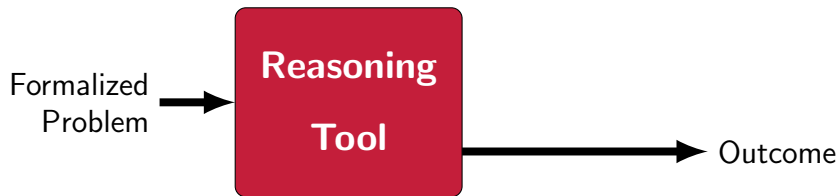
Simons Institute, 2023

<http://www.cs.cmu.edu/~bryant>

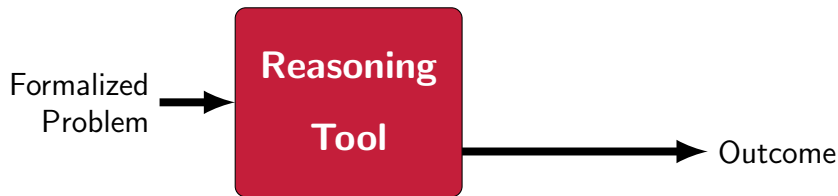
Background: My Research in “Formal” Verification

- 1990 *Formal* Verification of Digital Circuits Using Symbolic Ternary System Models
- 1991 *Formal* Hardware Verification by Symbolic Simulation
- 1991 *Formal* Verification of Memory Circuits by Switch-Level Simulation
- 1994 *Formally* Verifying a Microprocessor using a Simulation Methodology
- 1996 *Formal* Verification of PowerPC(TM) Arrays using Symbolic Trajectory Evaluation
- 1997 *Formal* Verification of a Superscalar Execution Unit
- 1998 *Formal* Verification of Pipelined Processors
- 1999 *Formal* Verification of an ARM Processor
- 2006 *Formal* Verification of Infinite State Systems Using Boolean Methods

Automated Reasoning Programs



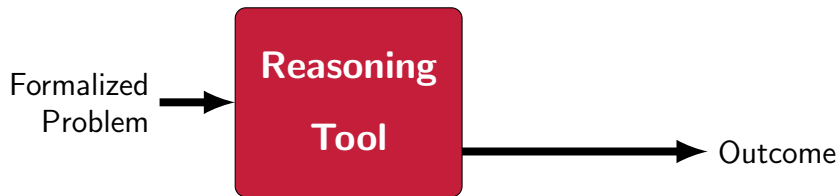
Automated Reasoning Programs



Standard Tools

- ▶ Lingering doubt about whether result can be trusted
- ▶ If find bug in tool, must rerun all prior verifications

Automated Reasoning Programs



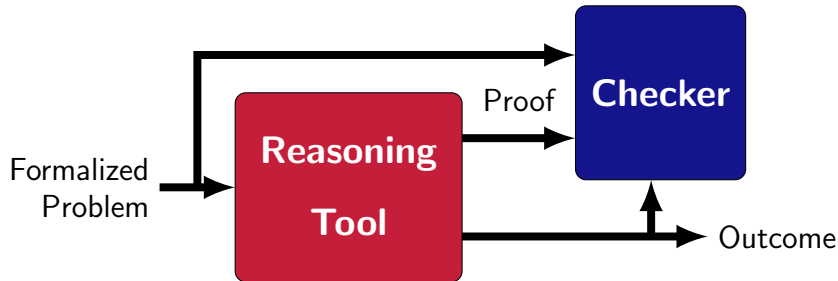
Standard Tools

- ▶ Lingering doubt about whether result can be trusted
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Formally Verified Tools

- ▶ Hard to develop
- ▶ Hard to make scalable

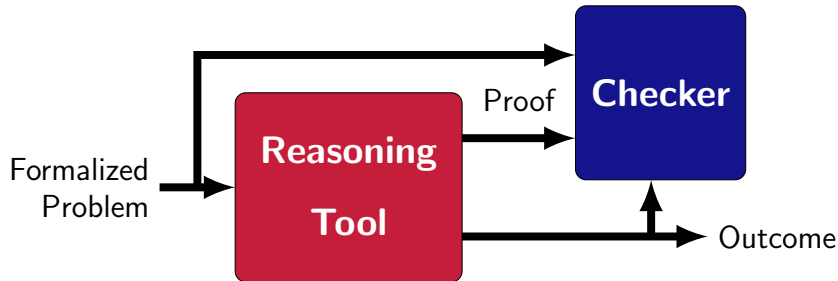
Proof-Generating Automated Reasoning Programs



Proof-Generating Tools

- ▶ Only need to prove individual executions, not entire program
- ▶ Can have bugs in tool but still trust result

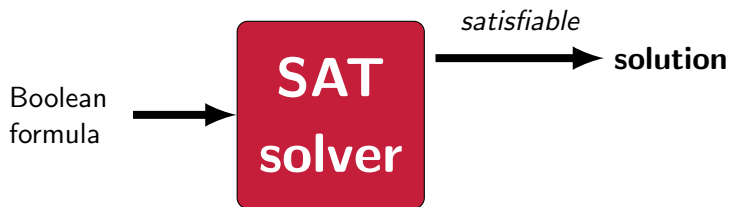
Proof-Generating Automated Reasoning Programs



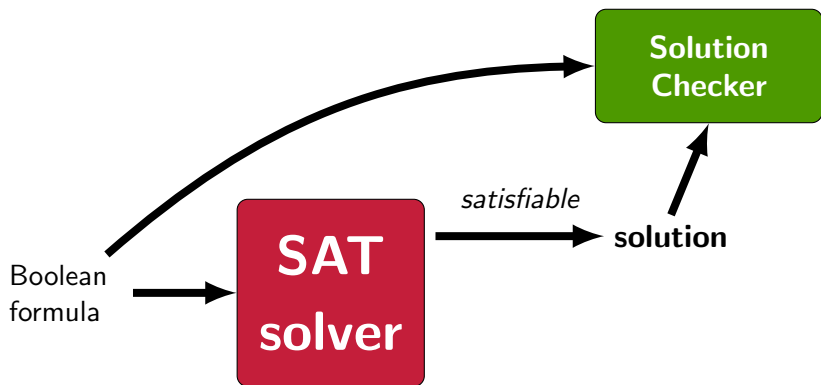
Proof-Generating Tools

- ▶ Only need to prove individual executions, not entire program
- ▶ Can have bugs in tool but still trust result
- ▶ Can we trust the checker?
 - Simple algorithms and implementation
 - Possibly formally verified

Boolean Satisfiability Solvers



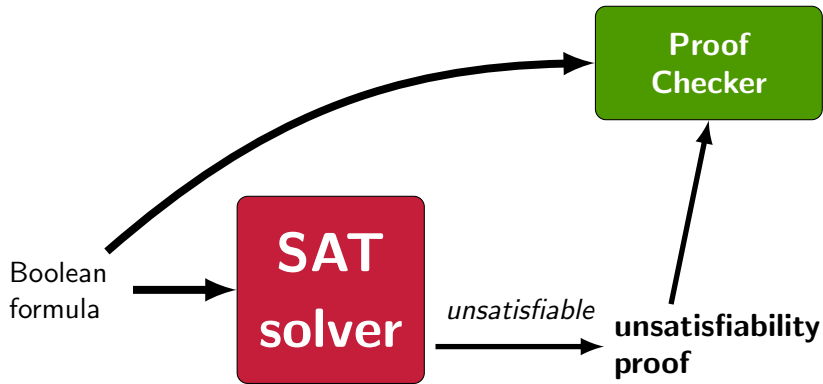
Boolean Satisfiability Solvers



Boolean Satisfiability Solvers



Proof Generating Solvers



Adoption of Proof Checking by SAT Community

Some History

- 2003 Proof generation added to zChaff [ZhaMal-2003] and BerkMin [GolNov-2003]
- 2013 Proof framework and checker well matched to CDCL solvers [HeuHunWet-2013]
- 2016 Proof generation mandatory for SAT competition main track

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Impact

- ▶ 2022 SAT Competition
 - No main track entrant reported UNSAT on satisfiable problem
 - Even on new benchmark formulas
- ▶ Tool developers alerted to bugs early in development
- ▶ Has enabled implementation of more complex & risky optimizations

Talk Overview

Add Proof Support to Other Forms of Automated Reasoning

Basics

- ▶ (Extended) resolution
- ▶ Clausal proofs

Binary Decision Diagrams (BDDs) and Proof Generation

- ▶ BDDs and extended resolution
- ▶ Supporting other Boolean reasoning methods

Certified Knowledge Compilation

- ▶ Partitioned-Operation Graphs (POGs)
- ▶ Equivalence proofs

Basics

Clauses

- ▶ $[\bar{u} \vee v \vee w]$ Disjunction of literals
- ▶ \perp Empty clause (False)

Resolution Principle

- ▶ Robinson, 1965

$$\frac{\bar{u} \vee v \vee w \quad \bar{w} \vee x \vee \bar{z}}{(\bar{u} \vee v) \vee (x \vee \bar{z})}$$

Basics

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$$(u \wedge \bar{v}) \rightarrow w \qquad w \rightarrow (x \vee \bar{z})$$

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$$(u \wedge \bar{v}) \rightarrow (x \vee \bar{z})$$

Clausal Proof Systems

Input Formula: Set of Clauses

$$C_1, C_2, \dots, C_m$$

Clausal Proof (of Unsatisfiability):

$$C_{m+1}, C_{m+2}, \dots, C_t$$

▶ Preserves satisfiability: For $i \geq m$:

if $\{C_1, C_2, \dots, C_i\}$ is satisfiable

then $\{C_1, C_2, \dots, C_i, C_{i+1}\}$ is satisfiable

▶ $C_t = \perp$

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- ▶ $C_t = \perp$
- ▶ Resolution rule preserves solutions:

$$\bigwedge_{1 \leq j \leq i} C_j \implies C_{i+1}$$

Clausal Proof Example

Step	Clause	Antecedents	Formula	
1	$[\bar{v} \vee w]$		$v \rightarrow w$	} Input clauses
2	$[\bar{v} \vee \bar{w}]$		$v \rightarrow \bar{w}$	
3	$[v]$		v	
4	$[\bar{v}]$	1, 2	\bar{v}	} Derived clauses
5	\perp	3, 4	$v \wedge \bar{v}$	

- ▶ Prove conjunction of input clauses unsatisfiable
- ▶ Add derived clauses
 - Provide list of antecedent clauses that resolve to new clause
- ▶ Finish with empty clause
 - Proof is series of inferences leading to contradiction

Extended Resolution

- ▶ Tseitin, 1967

Can introduce extension variables

- ▶ Variable e that has not yet occurred in proof
- ▶ Must introduce defining clauses
 - Clauses creating constraint of form $e \leftrightarrow F$
 - Boolean formula F over input and earlier extension variables

Extension variable becomes shorthand for larger formula

- ▶ Through repeated application, can have exponentially smaller proof

Extended Resolution Example

Example: Prove following set of constraints unsatisfiable

Constraint	Clauses
$u \wedge v \rightarrow w$	$[\bar{u} \vee \bar{v} \vee w]$
$u \wedge v \rightarrow \bar{w}$	$[\bar{u} \vee \bar{v} \vee \bar{w}]$
$u \wedge v$	$[u]$ $[v]$

- Strategy: Introduce extension variable e such that $e \leftrightarrow u \wedge v$

Constraint	Clauses
$u \wedge v \rightarrow e$	$[e \vee \bar{u} \vee \bar{v}]$
$e \rightarrow u$	$[\bar{e} \vee u]$
$e \rightarrow v$	$[\bar{e} \vee v]$

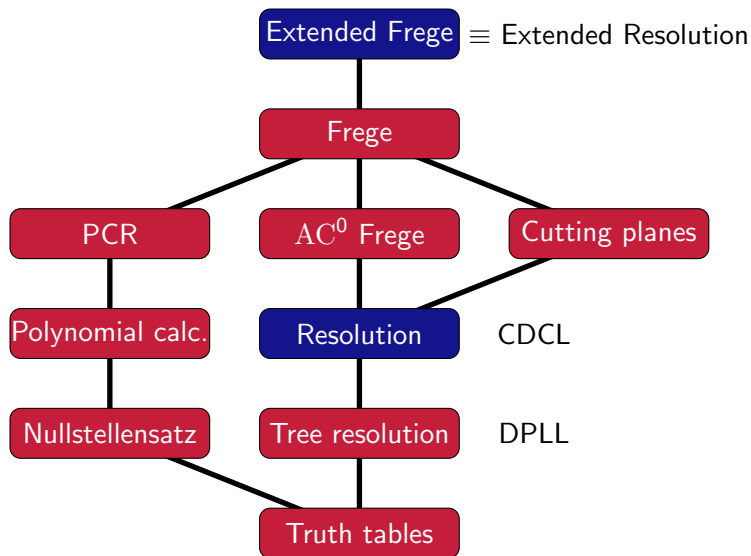
ER Proof

Step	Clause	Antecedents	Formula	
1	$[\bar{u} \vee \bar{v} \vee w]$		$u \wedge v \rightarrow w$	Input clauses
2	$[\bar{u} \vee \bar{v} \vee \bar{w}]$		$u \wedge v \rightarrow \bar{w}$	
3	$[u]$		u	
4	$[v]$		v	
5	$[e \vee \bar{u} \vee \bar{v}]$		$u \wedge v \rightarrow e$	Defining clauses
6	$[\bar{e} \vee u]$		$e \rightarrow u$	
7	$[\bar{e} \vee v]$		$e \rightarrow v$	
8	$[\bar{e} \vee \bar{v} \vee w]$	1, 6	$e \wedge v \rightarrow w$	Derived clauses
9	$[\bar{e} \vee w]$	7, 8	$e \rightarrow w$	
10	$[\bar{e} \vee \bar{v} \vee \bar{w}]$	2, 6	$e \wedge v \rightarrow \bar{w}$	
11	$[\bar{e} \vee \bar{w}]$	7, 10	$e \rightarrow \bar{w}$	
12	$[e \vee \bar{v}]$	3, 5	$v \rightarrow e$	
13	$[e]$	4, 12	e	
14	$[\bar{e}]$	9, 11	\bar{e}	
15	\perp	13, 14	$e \wedge \bar{e}$	

ER Proof

Step	Clause	Antecedents	Formula		
1	$[\bar{u} \vee \bar{v} \vee w]$		$u \wedge v \rightarrow w$	Input clauses	
2	$[\bar{u} \vee \bar{v} \vee \bar{w}]$		$u \wedge v \rightarrow \bar{w}$		
3	$[u]$		u		
4	$[v]$		v		
5	$[e \vee \bar{u} \vee \bar{v}]$		$u \wedge v \rightarrow e$	Defining clauses	
6	$[\bar{e} \vee u]$		$e \rightarrow u$		
7	$[\bar{e} \vee v]$		$e \rightarrow v$		
8	$[\bar{e} \vee \bar{v} \vee w]$	1, 6	$e \wedge v \rightarrow w$	Derived clauses	
9	$[\bar{e} \vee w]$	7, 8	$e \rightarrow w$		$u \wedge v$ replaced by e
10	$[\bar{e} \vee \bar{v} \vee \bar{w}]$	2, 6	$e \wedge v \rightarrow \bar{w}$		
11	$[\bar{e} \vee \bar{w}]$	7, 10	$e \rightarrow \bar{w}$		
12	$[e \vee \bar{v}]$	3, 5	$v \rightarrow e$		
13	$[e]$	4, 12	e		
14	$[\bar{e}]$	9, 11	\bar{e}		
15	\perp	13, 14	$e \wedge \bar{e}$		

Proof Complexity Hierarchy



The Power of (Extended) Resolution

Resolution

- ▶ Very weak

Implications for CDCL Solvers

- ▶ (Almost) every inference step can be expressed as polynomial number of resolution proof steps
- ▶ Exception: Bounded variable addition requires extension variables

Extended Resolution

- ▶ Can simulate all other known propositional proof systems
- ▶ No known class of formulas with superpolynomial lower bound

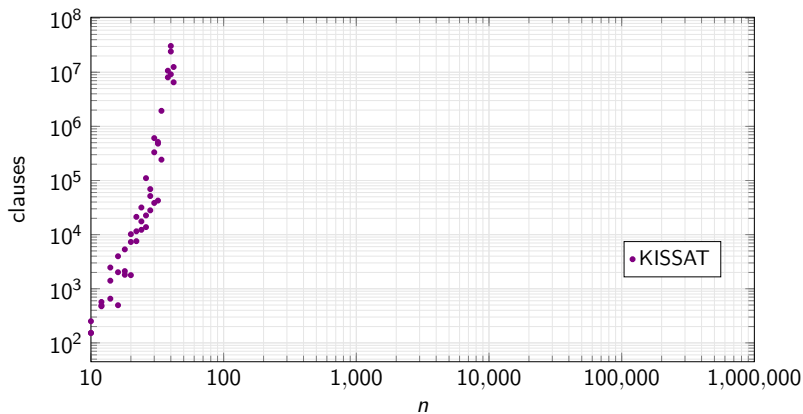
Parity Benchmark

- ▶ Chew and Heule, SAT 2020
- ▶ For random permutation π :

$$\begin{array}{l} x_1 \oplus x_2 \oplus \dots \oplus x_n = 1 \quad \text{Odd parity} \\ x_{\pi(1)} \oplus x_{\pi(2)} \oplus \dots \oplus x_{\pi(n)} = 0 \quad \text{Even parity} \end{array}$$

- ▶ Encode each equation in CNF
 - $n - 3$ auxiliary variables
 - Linear sequence of 3-argument parity constraints
- ▶ Conjunction unsatisfiable
- ▶ Very challenging for CDCL solvers

Chew-Heule Parity Benchmark Proof Sizes



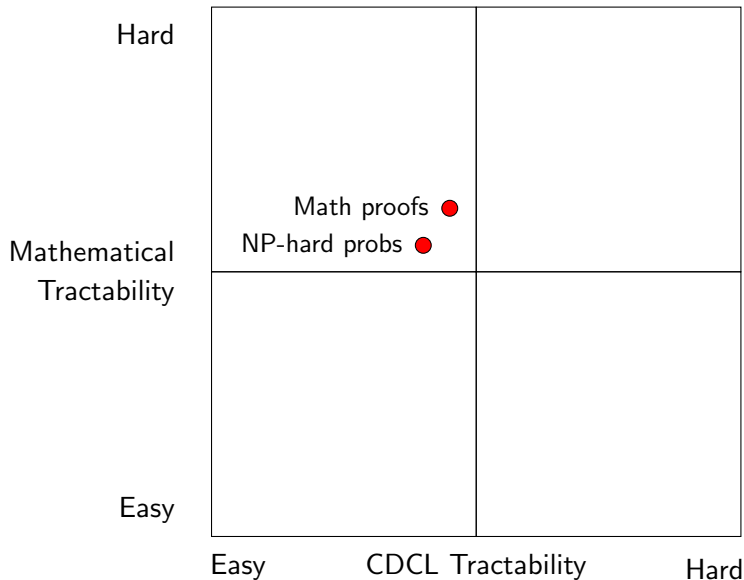
- ▶ KISSAT: State-of-the-art CDCL solver
- ▶ 3 different seeds for each value of n
- ▶ Cannot get beyond $n = 42$ within 600 seconds

A Perspective on the State of SAT Solving

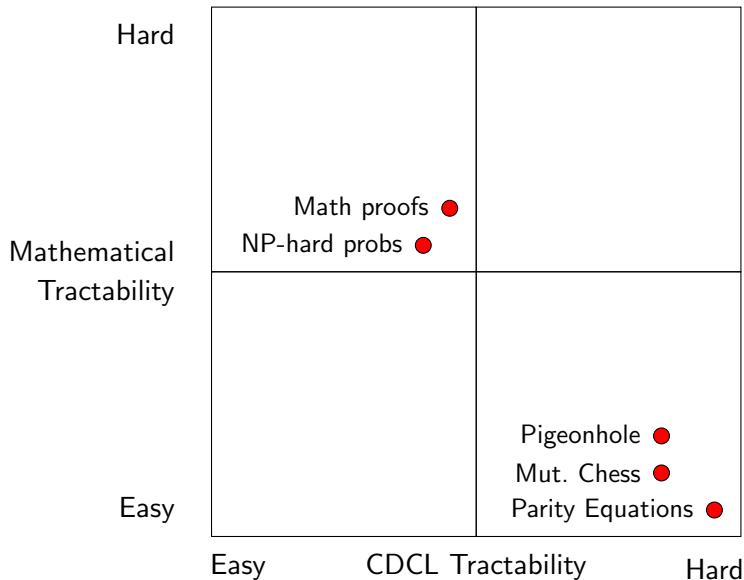
Hard		
Mathematical Tractability		
Easy		
	Easy	Hard

CDCL Tractability

A Perspective on the State of SAT Solving



A Perspective on the State of SAT Solving

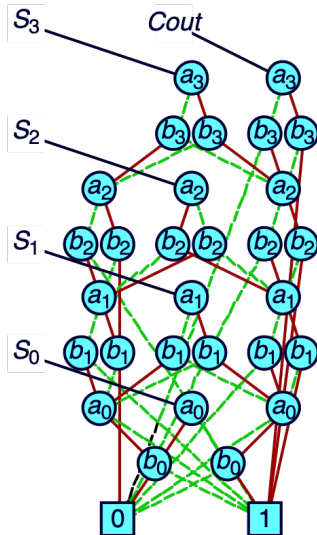


Reduced, Ordered Binary Decision Diagrams (BDDs)

- ▶ Bryant [Bry-1986]

Representation

- ▶ Canonical representation of Boolean function
- ▶ Compact for many useful cases



Proof-Generating SAT Solvers Based on BDDs

Implementations

- ▶ EBDDRES: Sinz, Biere, Jussila, 2006
[SinBie-2006, JusSinBie-2006]
- ▶ PGBDD: Bryant, Heule, 2021 [BryHeu-2021]
- ▶ TBUDDY: Bryant [Bry-2022]

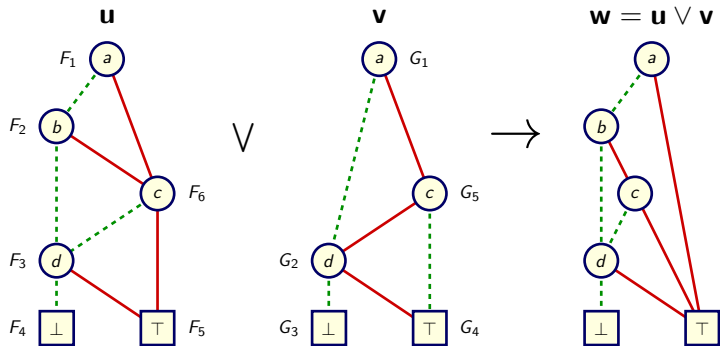
Extended-Resolution Proof Generation

- ▶ Introduce extension variable for each BDD node
- ▶ Proof steps based on recursive structure of BDD algorithms
- ▶ Proof is (very) detailed justification of each BDD operation

BDD Apply Algorithm

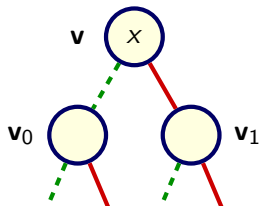
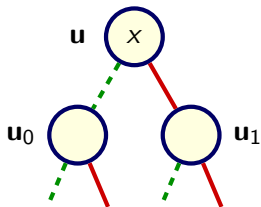
$$\mathbf{w} \leftarrow \mathbf{u} \odot \mathbf{v}$$

- ▶ $\mathbf{u}, \mathbf{v}, \mathbf{w}$ BDD root nodes representing Boolean functions
- ▶ \odot binary Boolean operator
 - ▶ E.g., \wedge, \vee, \oplus



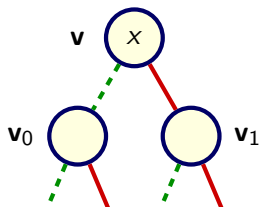
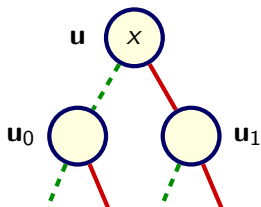
Apply Algorithm Recursion

Apply($\mathbf{u}, \mathbf{v}, \wedge$)

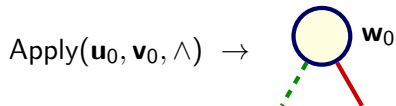
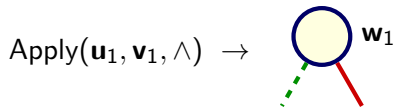


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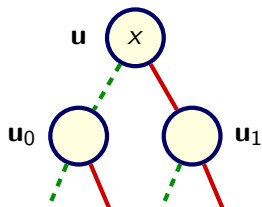


Recursion

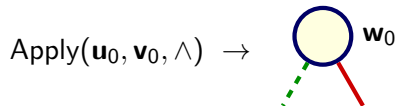
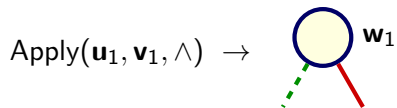


Apply Algorithm Recursion

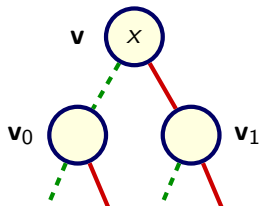
Apply($\mathbf{u}, \mathbf{v}, \wedge$)



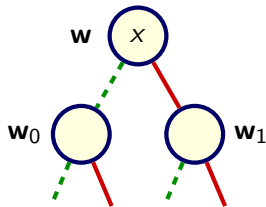
Recursion



Apply($\mathbf{v}, \mathbf{w}, \wedge$)

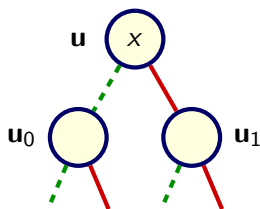


Result



Generating Extended Resolution Proofs

- ▶ Extension variable u for each node \mathbf{u} in BDD



- ▶ Defining clauses encode constraint $u \leftrightarrow ITE(x, u_1, u_0)$

Clause name	Formula	Clausal form
$HD(\mathbf{u})$	$x \rightarrow (u \rightarrow u_1)$	$[\bar{x} \vee \bar{u} \vee u_1]$
$LD(\mathbf{u})$	$\bar{x} \rightarrow (u \rightarrow u_0)$	$[x \vee \bar{u} \vee u_0]$
$HU(\mathbf{u})$	$x \rightarrow (u_1 \rightarrow u)$	$[\bar{x} \vee \bar{u}_1 \vee u]$
$LU(\mathbf{u})$	$\bar{x} \rightarrow (u_0 \rightarrow u)$	$[x \vee \bar{u}_0 \vee u]$

Proof-Generating Apply Operation

Integrate Proof Generation into Apply Operation

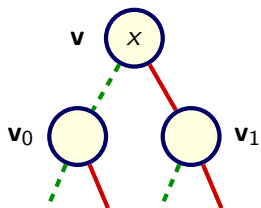
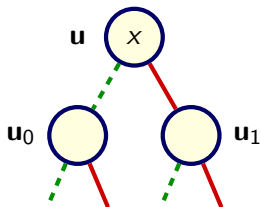
- ▶ $\text{Apply}(\mathbf{u}, \mathbf{v}, \wedge)$ returns \mathbf{w}
- ▶ Also generate proof $u \wedge v \rightarrow w$

Key Idea:

Proof follows recursion of the Apply algorithm

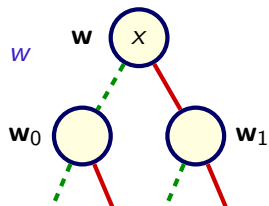
Apply Algorithm Recursion

Apply($\mathbf{u}, \mathbf{v}, \wedge$)



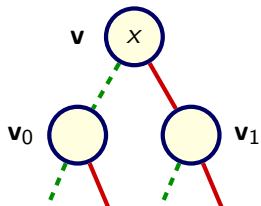
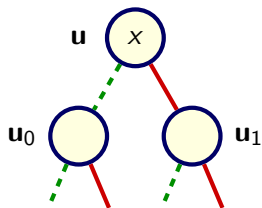
Result

$u \wedge v \rightarrow w$



Apply Algorithm Recursion

Apply($\mathbf{u}, \mathbf{v}, \wedge$)



Recursion

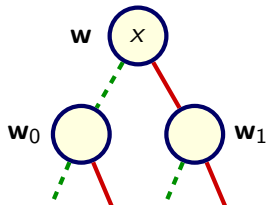
Apply($\mathbf{u}_1, \mathbf{v}_1, \wedge$) \rightarrow
 $u_1 \wedge v_1 \rightarrow w_1$



Apply($\mathbf{u}_0, \mathbf{v}_0, \wedge$) \rightarrow
 $u_0 \wedge v_0 \rightarrow w_0$



Result



Apply Proof Structure

Defining Clauses

Clause	Formula	Clause	Formula
HD(u)	$x \rightarrow (u \rightarrow u_1)$	LD(u)	$\bar{x} \rightarrow (u \rightarrow u_0)$
HD(v)	$x \rightarrow (v \rightarrow v_1)$	LD(v)	$\bar{x} \rightarrow (v \rightarrow v_0)$
HU(w)	$x \rightarrow (w_1 \rightarrow w)$	LU(w)	$\bar{x} \rightarrow (w_0 \rightarrow w)$

Resolution Steps

$$x \rightarrow (u \rightarrow u_1)$$

$$x \rightarrow (v \rightarrow v_1)$$

$$x \rightarrow (w_1 \rightarrow w) \quad u_1 \wedge v_1 \rightarrow w_1$$

$$x \rightarrow (u \wedge v \rightarrow w)$$

$$\bar{x} \rightarrow (u \rightarrow u_0)$$

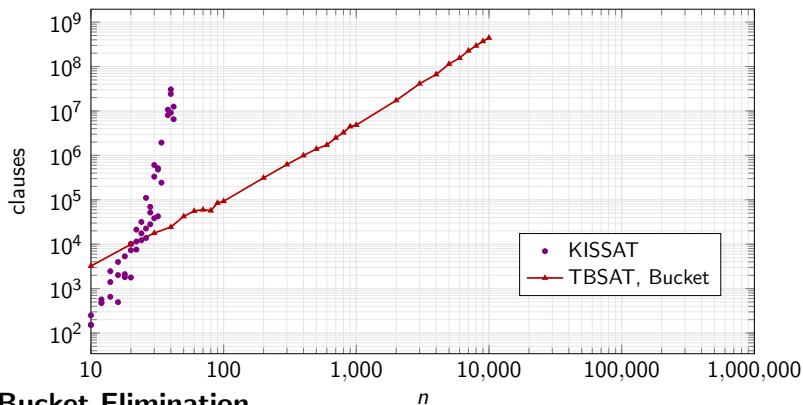
$$\bar{x} \rightarrow (v \rightarrow v_0)$$

$$\bar{x} \rightarrow (w_0 \rightarrow w) \quad u_0 \wedge v_0 \rightarrow w_0$$

$$\bar{x} \rightarrow (u \wedge v \rightarrow w)$$

$$u \wedge v \rightarrow w$$

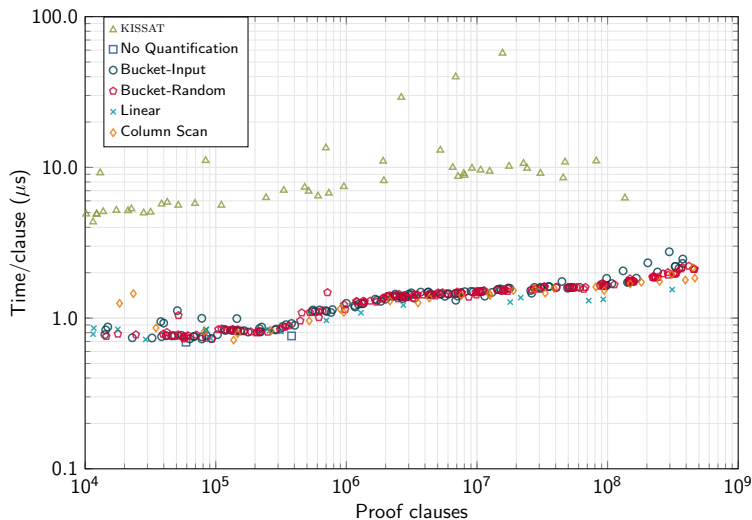
Chew-Heule Parity Benchmark Proof Sizes



Bucket Elimination

- ▶ Generate BDD representations of clauses
- ▶ Systematically form conjunctions and quantify out variables
 - Each recursive step generates up to 6 proof clauses
- ▶ Unsatisfiable formula generates BDD leaf node \perp

CDCL Proofs vs. BDD Proofs



- ▶ CDCL proof step indicates reduction in search space
- ▶ BDD proof steps justify algorithmic steps

Pseudo-Boolean (PB) Formulas

▶ Integer Equations

$$\sum_{1 \leq i \leq n} a_i x_i = b$$

- a_i, b : integer constants
- x_i : 0-1 valued variables

▶ Ordering Constraints

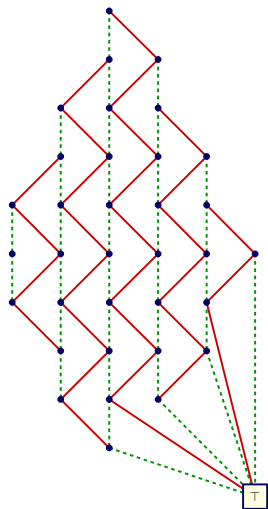
$$\sum_{1 \leq i \leq n} a_i x_i \geq b$$

▶ Modular Equations

$$\sum_{1 \leq i \leq n} a_i x_i \equiv b \pmod{r}$$

- r : constant modulus
- Parity constraint: $r = 2$

Representing PB Ordering Constraints with BDDs



- ▶ Example constraint:

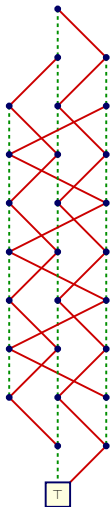
$$\begin{aligned} &+x_1 + x_3 + x_5 + x_7 + x_9 \\ &-x_2 - x_4 - x_6 - x_8 - x_{10} \end{aligned} \geq 0$$

- ▶ BDD size $\leq a_{\max} \cdot n^2$

$$a_{\max} = \max_{1 \leq i \leq n} |a_i|$$

- ▶ Independent of variable ordering

Representing PB Modular Equations with BDDs

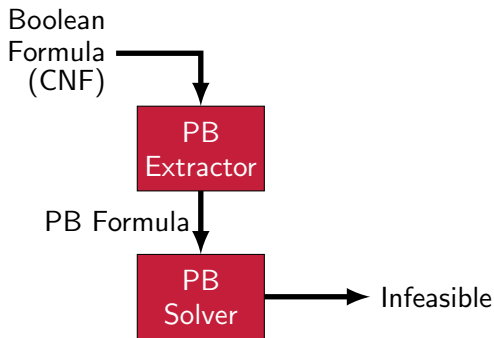


- ▶ Example equation:

$$\begin{aligned} +x_1 + x_3 + x_5 + x_7 + x_9 &\equiv 0 \pmod{3} \\ -x_2 - x_4 - x_6 - x_8 - x_{10} &\equiv 0 \pmod{3} \end{aligned}$$

- ▶ BDD size $\leq n \cdot r$
 - ▶ Independent of variable ordering

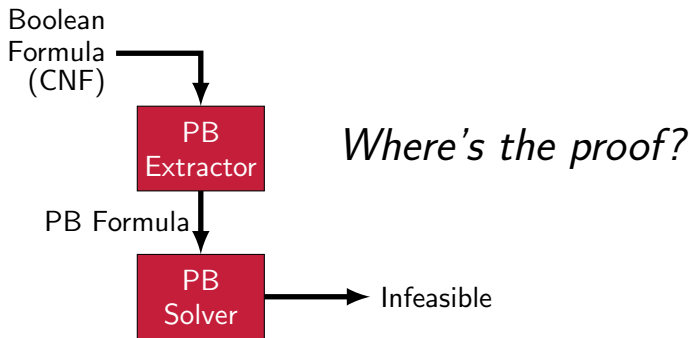
(Un)satisfiability with a Pseudo-Boolean Solver



Pseudo-Boolean Reasoning Methods

- ▶ (Modular) Equations
 - Gaussian elimination
- ▶ Ordering Constraints
 - Cutting planes
 - Fourier-Motzkin elimination

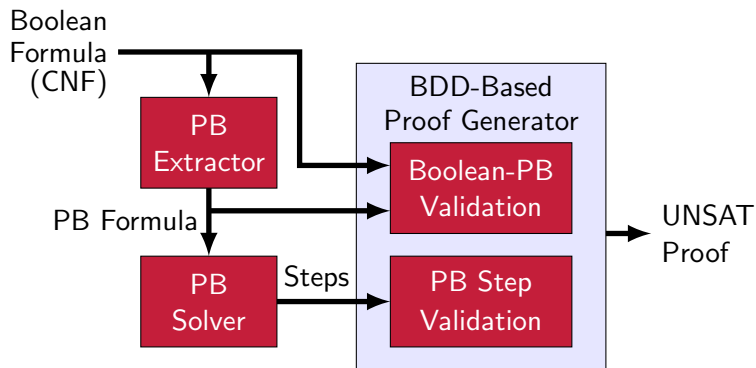
(Un)satisfiability with a Pseudo-Boolean Solver



Pseudo-Boolean Reasoning Methods

- ▶ (Modular) Equations
 - Gaussian elimination
- ▶ Ordering Constraints
 - Cutting planes
 - Fourier-Motzkin elimination

Integrating Pseudo-Boolean Reasoning into Proof-Generating SAT Solver [BryBieHeu-2022]



- ▶ Overall flow same as SAT solver
- ▶ PB solver does all of the reasoning
- ▶ BDDs serve only as mechanism for generating clausal proof

Validating Solver Steps

Individual Solver Step

- ▶ Given constraints p_i and p_j , compute new constraint p_k :

$$p_k \leftarrow p_i \odot p_j$$

- ▶ E.g., $\odot = +$

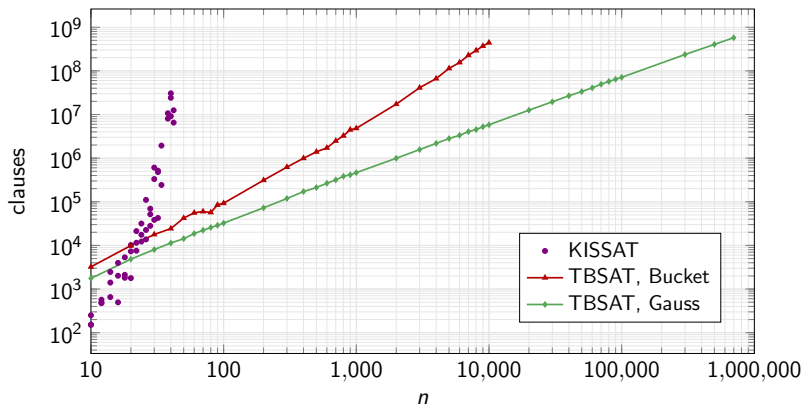
Validation

- ▶ Maintain $\text{BDD}(p)$ for each constraint p
- ▶ When generate p_k , also generate proof:

$$\text{BDD}(p_i) \wedge \text{BDD}(p_j) \implies \text{BDD}(p_k)$$

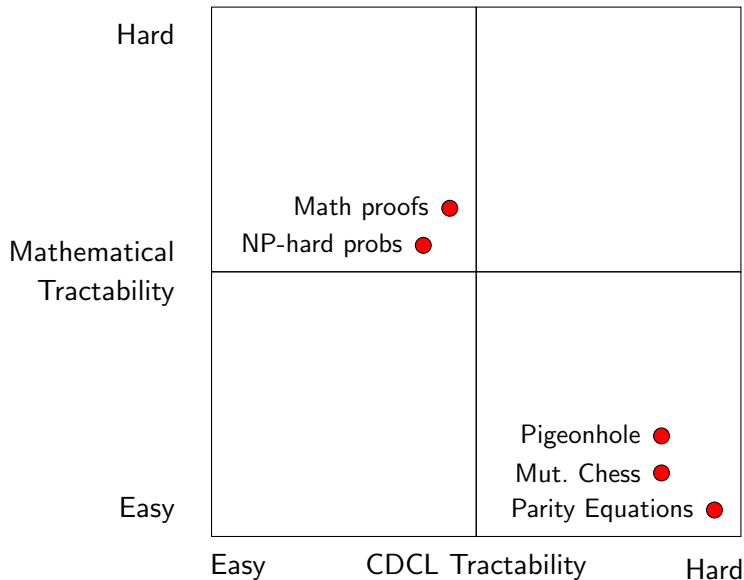
- ▶ Complexity $O(m_i \cdot m_j \cdot m_k)$
 - for BDDs of size m_i , m_j , and m_k

Chew-Heule Parity Benchmark Proof Sizes

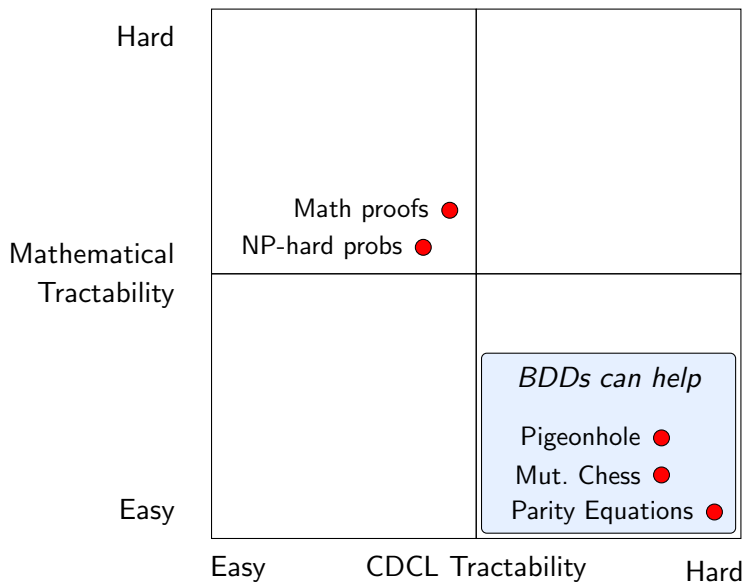


- ▶ Upper limit: $n = 699,051$
 - Node data structure sets limit of $2^{21} - 1$ BDD variables
 - CNF file has 2,097,147 variables and 5,592,392 clauses
- ▶ Some failures for large values of n due to poor pivot selection

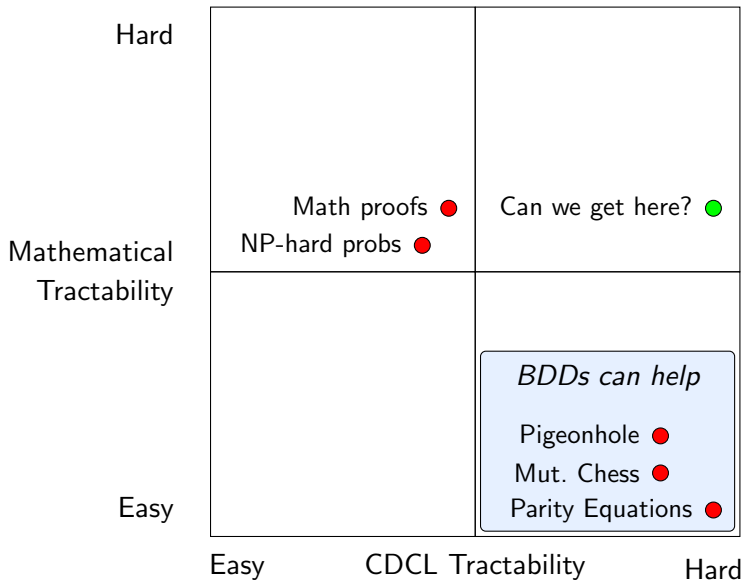
A Perspective on the State of SAT Solving



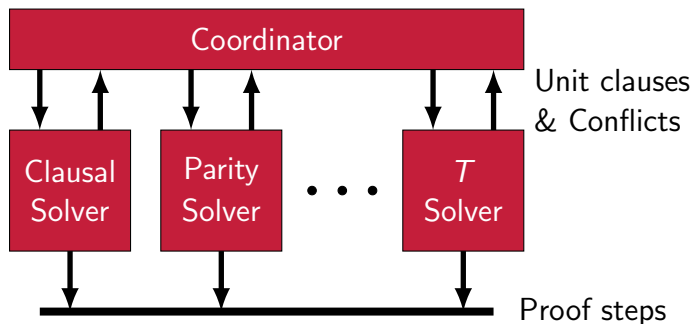
A Perspective on the State of SAT Solving



A Perspective on the State of SAT Solving



Proof Generation for CDCL(T)



- ▶ Solvers coordinate in unit propagation and conflict detection
- ▶ Proof generation: Each solver justifies its propagations & conflicts
- ▶ CryptoMiniSAT
 - Gauss-Jordan elimination for parity constraints
 - Can use BDDs to justify each elimination step [SooBry-22]

Knowledge Compilation

- ▶ Darwiche [DarMar-2002]

Convert CNF Formula into More Tractable Representation

Sample Query

- ▶ *Model Counting*: How many satisfying assignments does the formula have?

Challenging Problem

- ▶ #SAT more difficult than SAT

Knowledge Compilation

- ▶ Darwiche [DarMar-2002]

Convert CNF Formula into More Tractable Representation

Sample Query

- ▶ *Model Counting*: How many satisfying assignments does the formula have?

Challenging Problem

- ▶ #SAT more difficult than SAT

Questions

- ▶ *How do I know the generated representation is logically equivalent to the input formula?*
- ▶ *How do I know if the computed query values are correct?*

Algebraic Formulation

- ▶ Kimmig et al. [KimVdbDra-2017]

Definitions

- ▶ Input variables x_1, x_2, \dots, x_n
- ▶ *Assignment*: $\alpha = \{\ell_1, \ell_2, \dots, \ell_n\}$ with each $\ell_i \in \{x_i, \bar{x}_i\}$
- ▶ *Models*: $\mathcal{M}(\phi)$ is set of satisfying assignments for formula ϕ

Ring Evaluation

- ▶ Commutative ring \mathcal{R}
- ▶ Assign weight $w(x_i) \in \mathcal{R}$ to each input variable x_i
- ▶ Define $w(\bar{x}_i) = 1 - w(x_i)$
- ▶ Ring evaluation $\mathbf{R}(\phi, w)$ of formula ϕ :

$$\mathbf{R}(\phi, w) = \sum_{\alpha \in \mathcal{M}(\phi)} \prod_{\ell_i \in \alpha} w(\ell_i)$$

Ring Evaluation Examples

Model Counting

- ▶ Let $w(x_i) = w(\bar{x}_i) = 1/2$ for all i
- ▶ $\mathbf{R}(\phi, w)$ gives *density* of function
 - Fraction of assignments that satisfy ϕ
- ▶ Scale by 2^n to get model count

Probabilistic Inference

- ▶ Each input variable x_i is true with probability $p(x_i)$.
- ▶ $\mathbf{R}(\phi, p)$ is probability that formula is true

Partitioned-Operation Formulas

Allowed Operations

- ▶ **Product:** $\phi_1 \wedge^P \phi_2$, where $\mathcal{D}(\phi_1) \cap \mathcal{D}(\phi_2) = \emptyset$
 - $\mathcal{D}(\phi)$: Set of all variables occurring in ϕ
- ▶ **Sum:** $\phi_1 \vee^P \phi_2$, where $\mathcal{M}(\phi_1) \cap \mathcal{M}(\phi_2) = \emptyset$
- ▶ **Negation:** $\neg\phi$

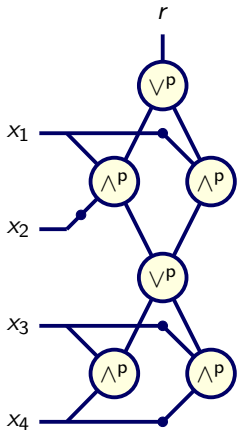
Ring Evaluation of Partitioned Formula

$$\mathbf{R}(\phi_1 \wedge^P \phi_2, w) = \mathbf{R}(\phi_1, w) \cdot \mathbf{R}(\phi_2, w)$$

$$\mathbf{R}(\phi_1 \vee^P \phi_2, w) = \mathbf{R}(\phi_1, w) + \mathbf{R}(\phi_2, w)$$

$$\mathbf{R}(\neg\phi, w) = 1 - \mathbf{R}(\phi, w)$$

Partitioned-Operation Graphs (POGs)



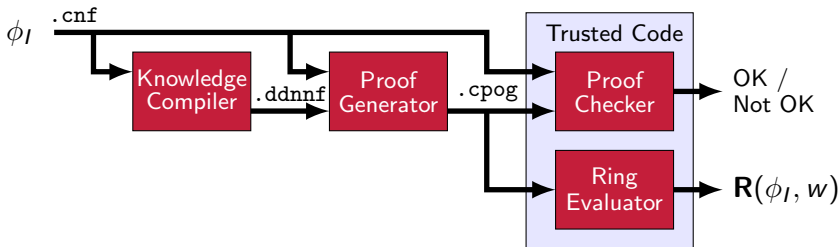
► Directed graph representation of formula

- Leaf nodes: Input variables
- Operation nodes: Partitioned product and sum
- Each edge can be negated

► Can encode other compiled representations

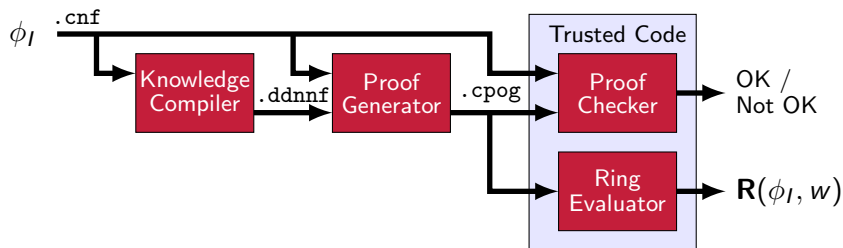
Certifying Toolchain

- Joint work with Wojciech Nawrocki, Jeremy Avigad, and Marijn Heule



- ▶ *Knowledge Compiler* (D4 [LagMar-2017]): Convert CNF into representation using only partitioned operations
- ▶ *Proof Generator*: Generate file combining POG definition + equivalence proof
- ▶ *Proof Checker*: Validate proof file
- ▶ *Ring Evaluator*: Compute standard or weighted model count

Trusting the Trusted Code



Within Lean 4 Proof Framework [DemUlr-2021]

- ▶ Soundness of proof system
 - Helped us identify some weaknesses in our proof rules
- ▶ Verified checker
 - Around $6\times$ slower than one implemented in C
- ▶ Ring Evaluator: Over rationals

CPOG Declaration + Proof

Input Formula ϕ_I

POG Declaration θ_P

- ▶ Extension variable u for each operation node \mathbf{u}
- ▶ Node \mathbf{u} with k children characterized by $k + 1$ defining clauses
- ▶ Children indicated by literals
 - Positive or negated arguments
 - Input variables or results from other operation
- ▶ Unit clause $[r]$ for root node \mathbf{r}

Proof Objective

$$\phi_I \iff \theta_P$$

CPOG Proof Structure

Forward Implication Proof

$$\phi_I \implies \theta_P$$

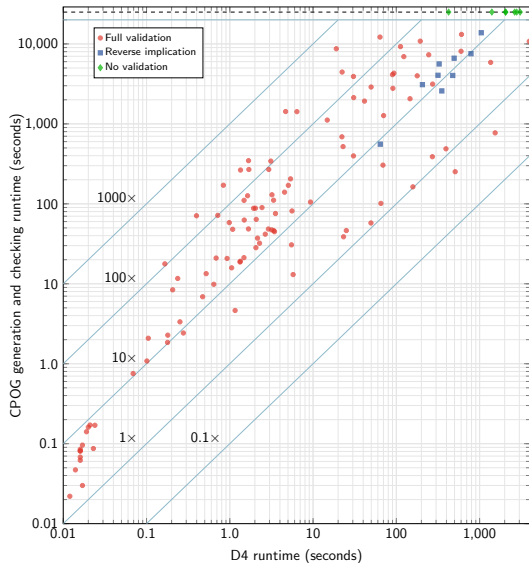
- ▶ Add clauses by resolution
- ▶ Terminating with unit clause $[r]$
- ▶ *Any assignment satisfying ϕ_I causes the POG formula to evaluate to true*

Reverse Implication Proof

$$\theta_P \implies \phi_I$$

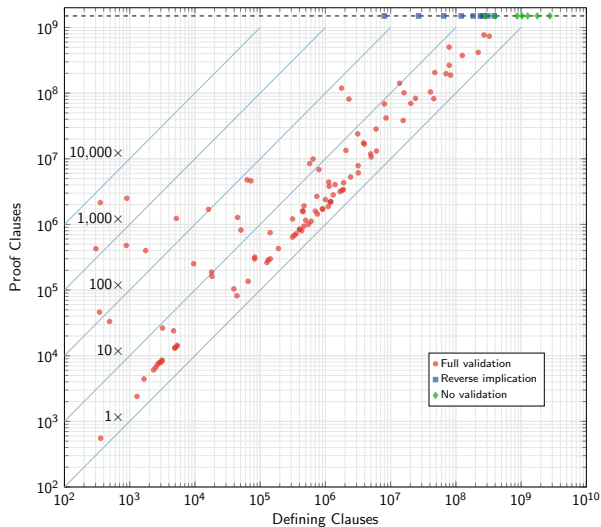
- ▶ Delete clauses by resolution
 - Deleted clause implied by remaining ones
- ▶ Including each of the input clauses
- ▶ *Any assignment falsifying the input clause causes the POG formula to evaluate to false*

Experimental Results: CPOG Generation and Checking



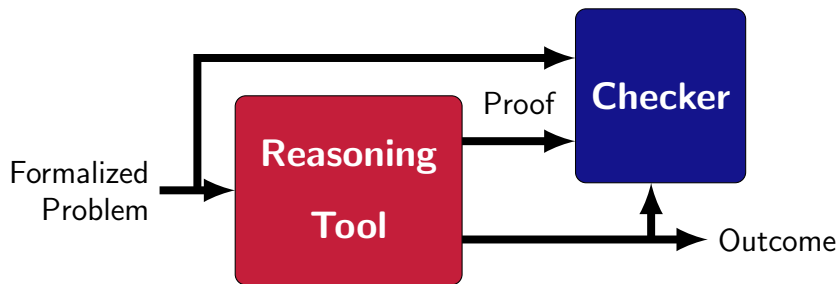
- 180 benchmark files from 2022 model checking competitions
- D4 completed 124 with 4000-second time limit
- Generated complete proofs for 108 with 10,000-second time limit
- Reverse implications for 9
- No proofs for 7

Experimental Results: CPOG Sizes



- 108 problems fully verified
- CPOG files up to 160 GB
- Reverse implications for 9
- No proofs for 7

Recap: Important Principle



Proof-Generating Tools

- ▶ Formally verifying a large, complex program is impractical
- ▶ Instead, certify individual executions of the program

Important Concepts

Checkable Proofs of Program Executions

- ▶ Very habit forming
- ▶ Many research possibilities

Clausal Proof Frameworks

- ▶ Well understood set of principles
- ▶ Can build on existing infrastructure
 - E.g., our CPOG proof generator uses CaDiCal and Drat-trim
- ▶ Not just for refutation proofs

Extended Resolution

- ▶ Can reason about other representations of Boolean formulas
- ▶ Can introduce intermediate proof structures
 - E.g., CPOG proof generator uses lemmas to control recursion
 - Validation for each shared POG node generated once and used multiple times

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