

Trustworthy Boolean Reasoning 2: BDDs and SAT

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Important Ideas for These Lectures

- ▶ SAT solvers are useful tools
 - ▶ Many practical problems reducible to SAT
 - ▶ Need to learn effective encoding techniques
- ▶ For many applications, formulas should be unsatisfiable
 - ▶ Program should generate a checkable proof
 - ▶ There is a well-developed proof infrastructure
- ▶ **Binary Decision Diagrams (BDDs) can play important role**
 - ▶ **In supplementing current SAT algorithms**
 - ▶ **In proof generation**

Reduced Ordered Binary Decision Diagrams (BDDs)

- ▶ Bryant, 1986
- ▶ Based on earlier work by Lee (1959) and Akers (1978)

Graph Representation of Boolean Functions

- ▶ Canonical Form
- ▶ Compact for many useful problems
- ▶ Simple algorithms to construct & manipulate

Used in SAT, Model Checking, ...

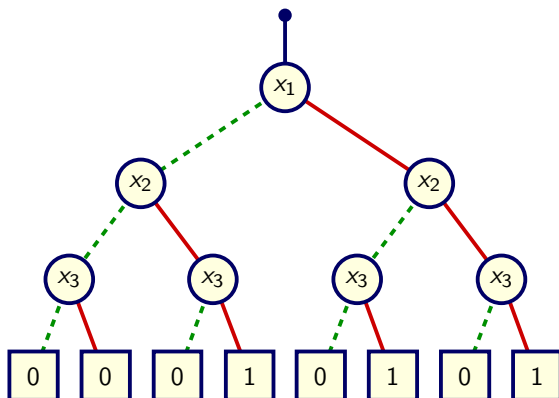
- ▶ Bottom-up approach
 - ▶ Construct canonical representation of problem
 - ▶ Generate solutions
- ▶ Compare to search-based methods
 - ▶ E.g., CDCL
 - ▶ Top-down approaches
 - ▶ Keep branching on variables until find solution

Boolean Function Representations

Truth Table

x_1	x_2	x_3	f
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	0
1	1	1	1

Decision Tree



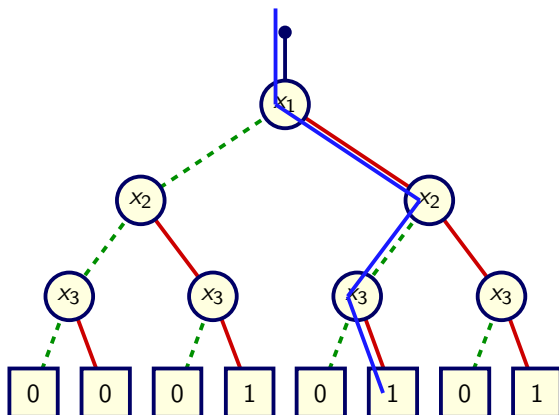
► Size = $O(2^n)$

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Decision Tree



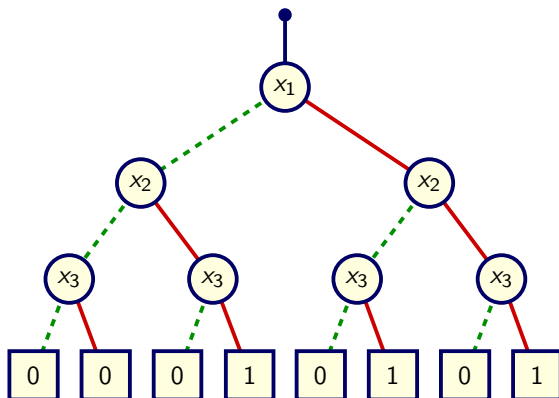
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Reducing to Canonical Form

Truth Table

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Graph Representation



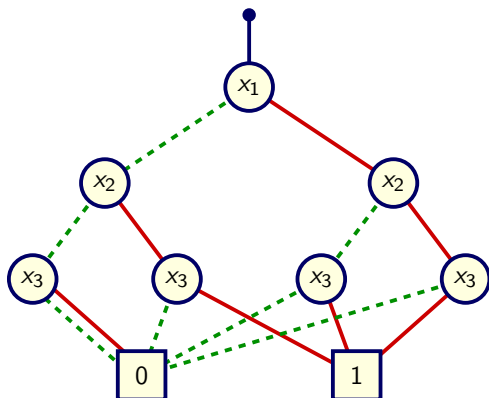
- ▶ Merge isomorphic nodes
- ▶ Eliminate redundant tests

Reducing to Canonical Form

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Graph Representation



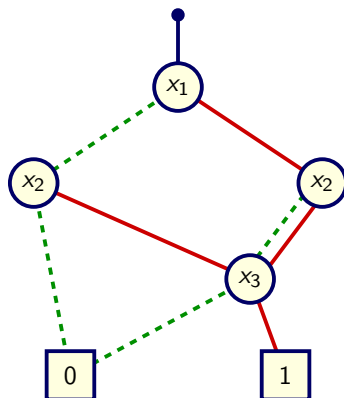
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Reducing to Canonical Form

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Graph Representation



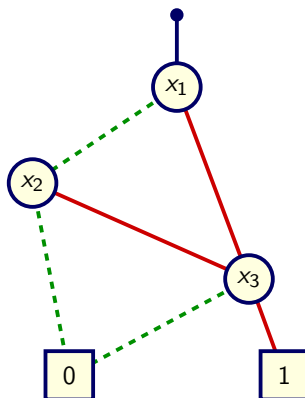
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Reducing to Canonical Form

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Graph Representation



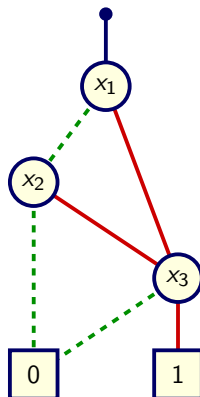
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Canonical Form

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1	1	1	1

Reduced Ordered
Binary Decision Diagram



- ▶ Canonical representation of Boolean function
- ▶ No further simplifications possible

BDD Representation of Unsatisfiable Formula

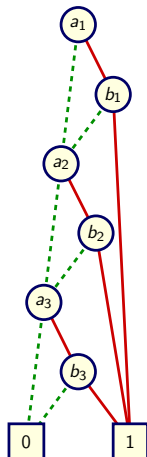
0

- ▶ Refer to this as \perp
- ▶ Unique
- ▶ Converting from CNF to BDD may require exponential number of steps

Effect of Variable Ordering

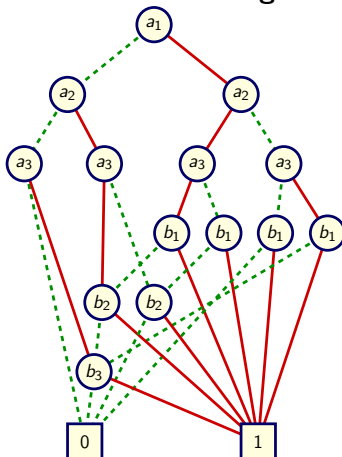
$$(a_1 \wedge b_1) \vee (a_2 \wedge b_2) \vee (a_3 \wedge b_3)$$

Good Ordering



► Linear growth

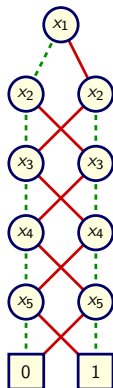
Bad Ordering



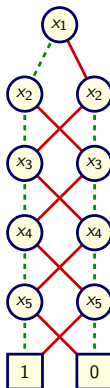
► Exponential growth

BDD Representation of Parity Constraints

Odd Parity



Even Parity



- ▶ Linear complexity
- ▶ Insensitive to variable order
- ▶ Potential major advantage over CDCL

Symbolic Manipulation with BDDs

Strategy

- ▶ Represent data as set of BDDs
 - ▶ All with same variable ordering
- ▶ Express method as sequence of symbolic operations
 - ▶ Generate new BDDs. Test properties of BDDs
- ▶ Implement each operation via BDD manipulation
 - ▶ Never enumerate individual cases
 - ▶ Efficient, as long as BDDs stay small

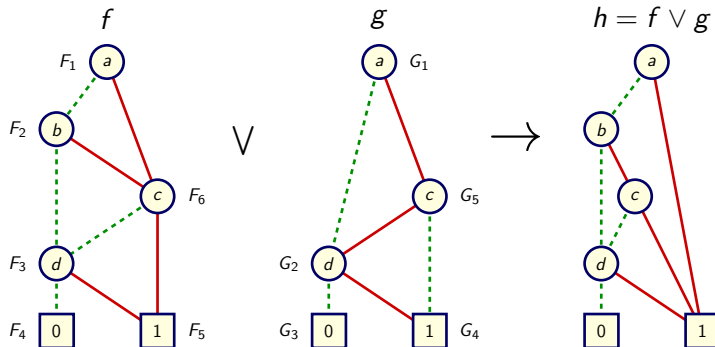
Key Algorithmic Properties

- ▶ Arguments at each step are BDDs with same variable ordering
- ▶ Result is BDD with same ordering
- ▶ Each step has polynomial complexity

Apply Algorithm

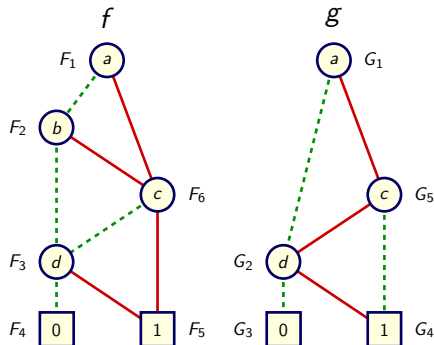
$$h \leftarrow f \odot g$$

- ▶ f, g, h functions represented as BDDs
- ▶ \odot binary Boolean operator
 - ▶ E.g., \wedge, \vee, \oplus

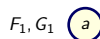


Apply Algorithm Recursion

- ▶ Recurse through argument graphs
- ▶ Stop when hit terminal case
- ▶ Save results in cache to reuse when hit same arguments

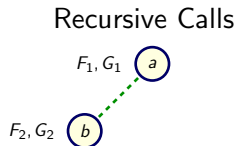
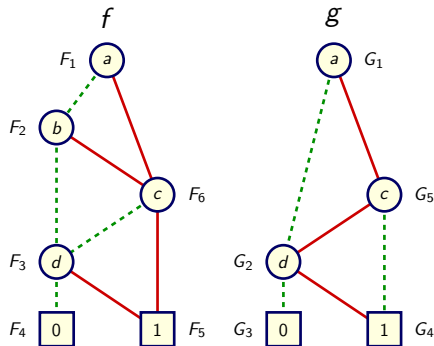


Recursive Calls



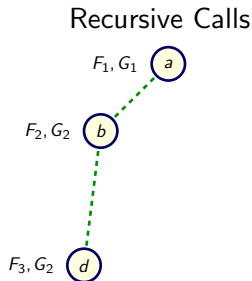
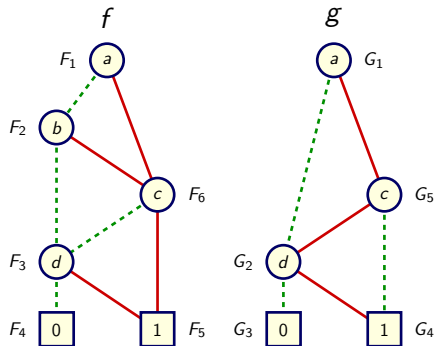
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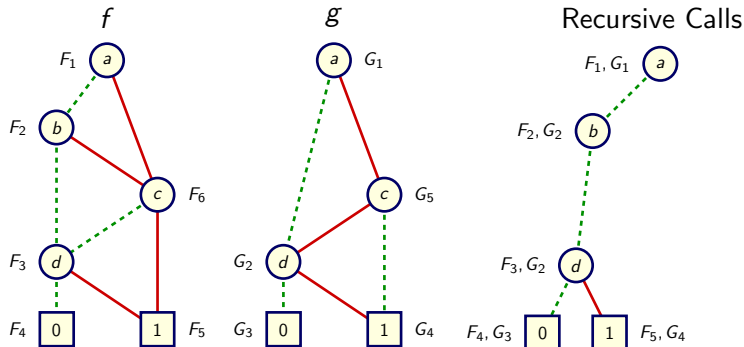
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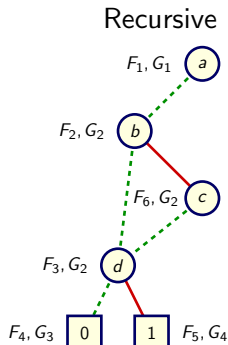
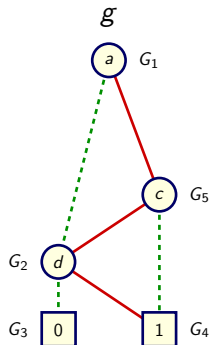
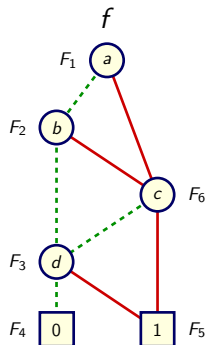
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Apply Algorithm Recursion

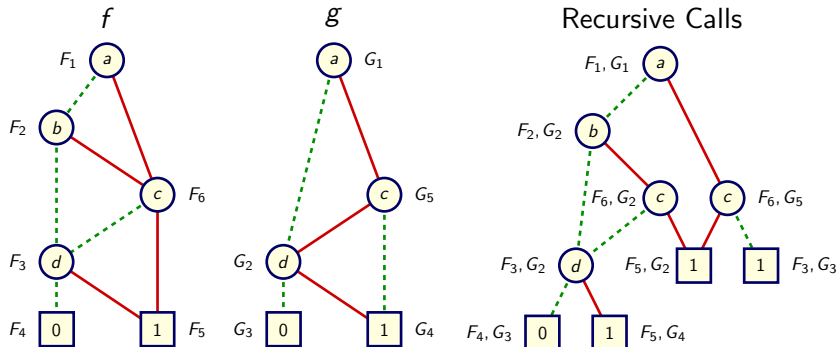
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Recursive Calls

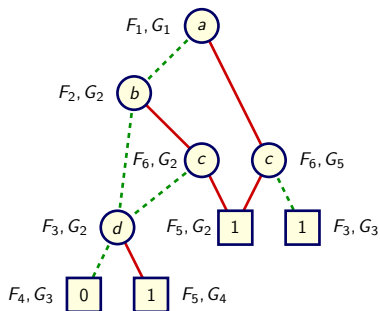
Apply Algorithm Recursion

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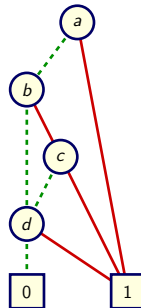


Apply Algorithm Result

Recursive Calls



Reduced Result



BDD-Based SAT Solving: Direct Evaluation

Algorithm

1. Compute BDD t_i for each input clause C_i
2. Form conjunction

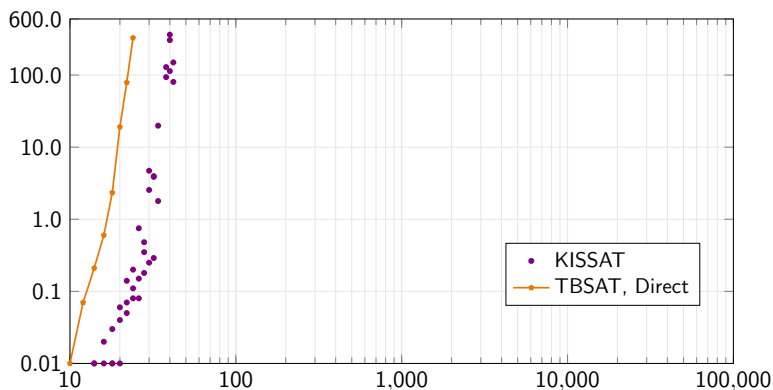
$$s = \bigwedge_{1 \leq i \leq m} t_i$$

- ▶ E.g., with linear or tree evaluation
3. Return UNSAT ($s = \perp$) or SAT ($s \neq \perp$)

Practicality

- ▶ Only for small problems
- ▶ Resulting BDD s represents *all* solutions

Parity Benchmark Runtime



- ▶ TBSAT: BDD-Based SAT Solver
- ▶ In direct mode, even worse than KISSAT
- ▶ Limited to $n \leq 24$ within 600 seconds

BDD-Based SAT Solving: Bucket Elimination

- ▶ Maintain list (“bucket”) B_j for each variable x_j
- ▶ Each BDD stored in bucket according to root node variable

Algorithm

Initialization:

Form BDD t_i for each input clause C_i

Place each t_i in bucket according to $Var(t_i)$

For each bucket B_j :

Form conjunction s_j of all BDDs in bucket B_j

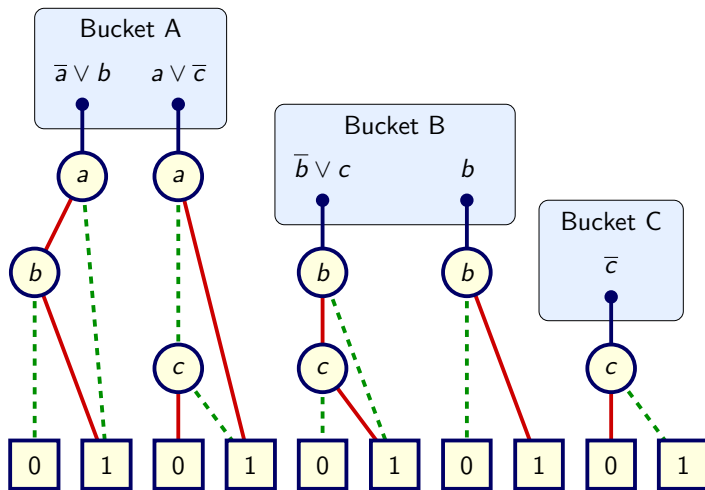
If $s_j = \perp$ then return UNSAT

Compute $r_j = \exists x_j s_j$

Place r_j in bucket according to $Var(r_j)$

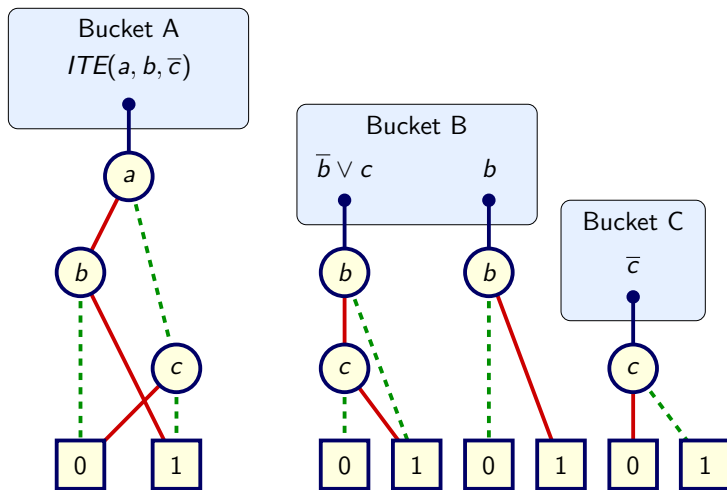
return SAT

Bucket Elimination Example



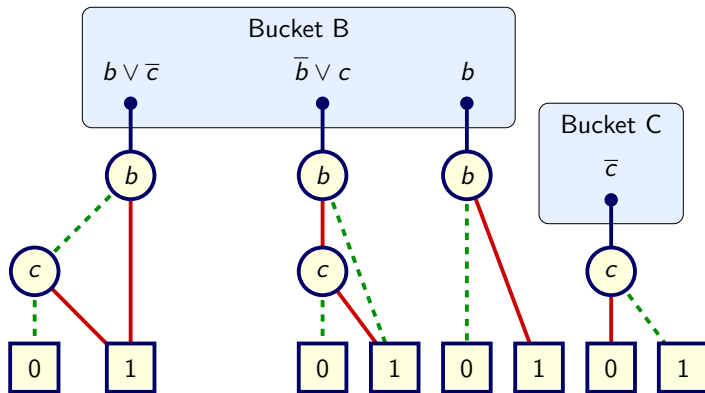
- Initially: BDD for each input clause
- In bucket according to root variable

Bucket Elimination Example



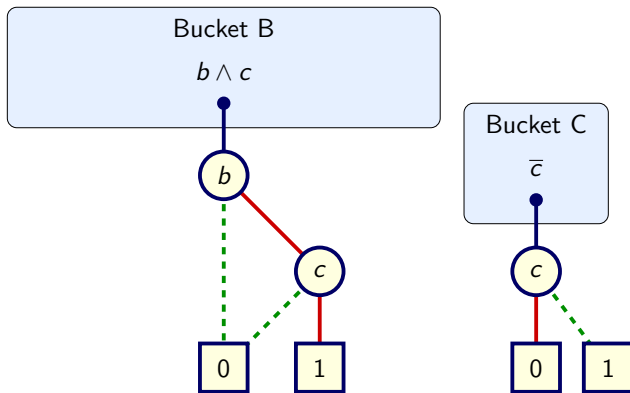
- Conjunct BDDs in topmost bucket A

Bucket Elimination Example



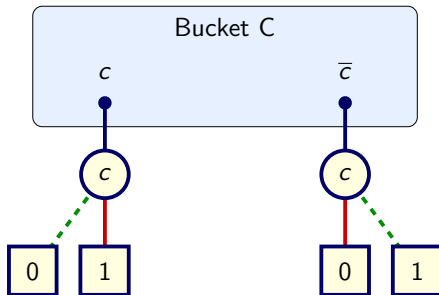
- Existentially quantify variable a
- Place result in appropriate bucket

Bucket Elimination Example



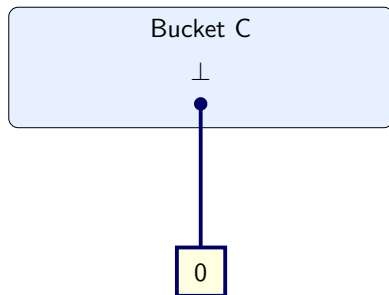
- Conjunct BDDs in bucket B

Bucket Elimination Example



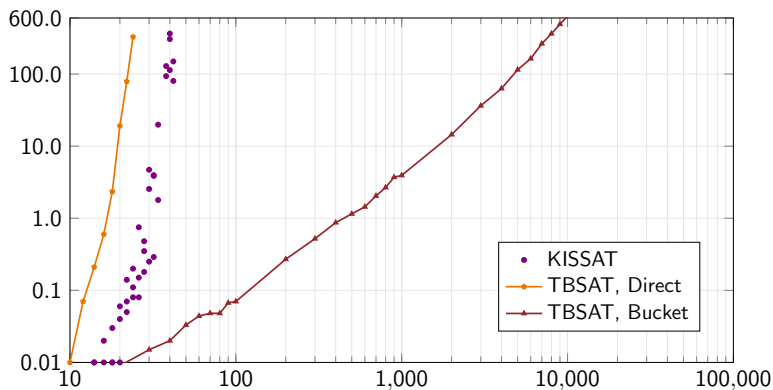
- ▶ Existentially quantify variable b
- ▶ Place result in appropriate bucket

Bucket Elimination Example



- ▶ Conjoin BDDs in bucket C
- ▶ Final result will be \perp or \top

Parity Benchmark Runtime



- ▶ $n = 10,000$ in 633 seconds
- ▶ Large benefit from quantification
 - ▶ abstracts away intermediate variables

What Parity Benchmark Demonstrates

- ▶ Binary Decision Diagrams (BDDs) can play important role in SAT
 - ▶ In supplementing current SAT algorithms
 - ▶ But, *what about proof generation?*

Extended Resolution and BDDs

- ▶ Tseitin, 1967

Can introduce extension variables

- ▶ Variable z that has not yet occurred in proof
- ▶ Must add *defining* clauses
 - ▶ Encode constraint of form $z \leftrightarrow F$
 - ▶ Boolean formula z over input and earlier extension variables

Extension variable z becomes shorthand for formula F

- ▶ Repeated use can yield exponentially smaller proof

Extended Resolution and BDDs

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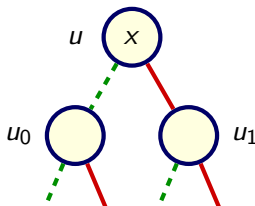
- ▶ Repeated use can yield exponentially smaller proof

Generate extension variable for every node in BDD

- ▶ Biere, Sinz, Jussila, 2006
- ▶ Each recursive step of Apply algorithm justified as proof steps
- ▶ Reducing formula to BDD \perp yields UNSAT proof

Generating Extended Resolution Proofs

- ▶ Create extension variable for each node in BDD
 - ▶ Notation: Same symbol for node and its extension variable

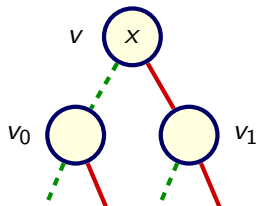
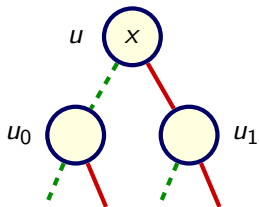


- ▶ Defining clauses encode constraint $u \leftrightarrow \text{ITE}(x, u_1, u_0)$

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

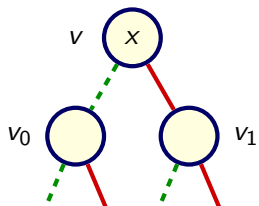
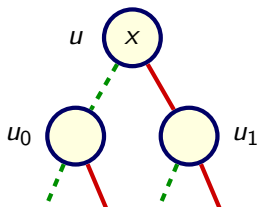
Apply Algorithm Recursion

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

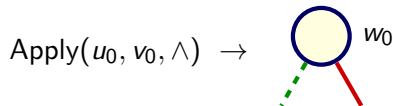
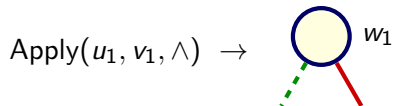


Apply Algorithm Recursion

Apply(u, v, \wedge)

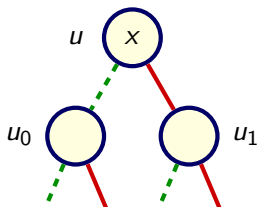


Recursion

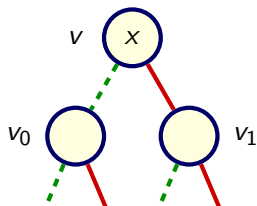
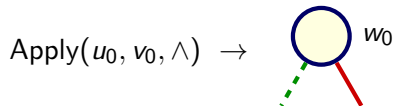
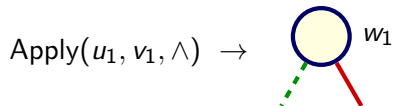


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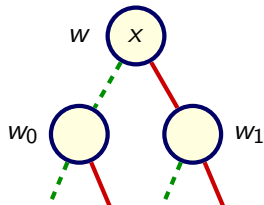
Apply(u, v, \wedge)



Recursion



Result



Proof-Generating Apply Operation

Integrate Proof Generation into Apply Operation

- ▶ When $\text{Apply}(u, v, \wedge)$ returns w , also generate proof $u \wedge v \rightarrow w$
- ▶ **Key Idea:** Proof based on the underlying logic of the Apply algorithm

Proof Structure

- ▶ Assume recursive calls generate proofs
 - ▶ $u_1 \wedge v_1 \rightarrow w_1$
 - ▶ $u_0 \wedge v_0 \rightarrow w_0$
- ▶ Combine with defining clauses for nodes u , v , and w

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

$$\begin{array}{c} x \rightarrow (u \rightarrow u_1) \\ x \rightarrow (v \rightarrow v_1) \\ x \rightarrow (w_1 \rightarrow w) \quad u_1 \wedge v_1 \rightarrow w_1 \\ \hline x \rightarrow (u \wedge v \rightarrow w) \end{array} \qquad \begin{array}{c} \bar{x} \rightarrow (u \rightarrow u_0) \\ \bar{x} \rightarrow (v \rightarrow v_0) \\ \bar{x} \rightarrow (w_0 \rightarrow w) \quad u_0 \wedge v_0 \rightarrow w_0 \\ \hline \bar{x} \rightarrow (u \wedge v \rightarrow w) \end{array}$$
$$\hline u \wedge v \rightarrow w$$

Can perform with 2 RUP steps

Quantification Operations

Operation $\text{EQuant}(f, X)$

- ▶ Abstract away details of satisfying (partial) solutions
- ▶ Not logically required for SAT solver
 - ▶ But, critical for obtaining good performance

Proof Generation

- ▶ Do not attempt to follow recursive structure of algorithm
- ▶ Instead, follow with separate implication proof generation
 - ▶ $\text{EQuant}(u, X) \rightarrow w$
 - ▶ Generate proof $u \rightarrow w$
 - ▶ Algorithm similar to proof-generating Apply operation

Overall Proof Task

Input Variables

Input Clauses

- ▶ Set of input clauses C_I over the input variables

Completion

- ▶ Generate Proof $C_I \models \perp$

Trusted BDDs (TBDD)

Components

- ▶ BDD with root node t
- ▶ Proof step for unit clause (t)

Interpretation

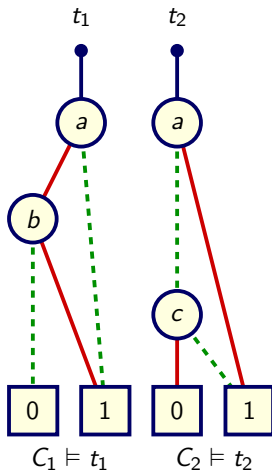
- ▶ $C_I \models t$
- ▶ Any variable assignment that satisfies input clauses must yield 1 for BDD with root t

TBDD Example

$$\begin{array}{ll} C_1 & \bar{a} \vee b \\ C_2 & a \vee \bar{c} \end{array}$$

$$t_1 \longleftarrow \text{FromClause}(C_1)$$

$$t_2 \longleftarrow \text{FromClause}(C_2)$$



TBDD Example

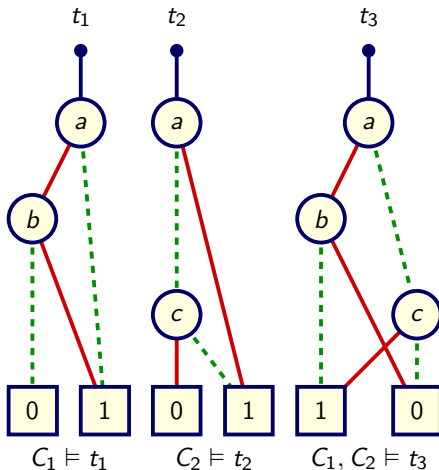
$$C_1 \quad \bar{a} \vee b$$

$$C_2 \quad a \vee \bar{c}$$

$$t_1 \longleftarrow \text{FromClause}(C_1)$$

$$t_2 \longleftarrow \text{FromClause}(C_2)$$

$$t_3 \longleftarrow \text{ApplyAnd}(t_1, t_2)$$



Structure of Overall Proof

Input Variables

- ▶ BDD variable for each input variable

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Input Clauses

- ▶ For each input clause $C_i \in C_I$, generate BDD representation t_i
- ▶ Generate *validation* proof $C_i \models t_i$
 - ▶ Sequence of resolution steps based on linear structure of BDD
- ▶ Initial set of TBDDs

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Combine Top-Level BDDs

- ▶ Choose TBDDs t_i, t_j . Use to generate TBDD t_k
- ▶ $t_k \leftarrow \text{ApplyAnd}(t_i, t_j)$
 - ▶ Combine proofs $C_I \models t_i, C_I \models t_j$ and $t_i \wedge t_j \rightarrow t_k$ to validate $C_I \models t_k$
- ▶ $t_k \leftarrow \text{EQuant}(t_i, X)$
 - ▶ Combine proofs $C_I \models t_i$ and $t_i \rightarrow t_k$ to validate $C_I \models t_k$

Structure of Overall Proof

Input Variables

- ▶ BDD variable for each input variable

Input Clauses

- ▶ For each input clause $C_i \in C_I$, generate BDD representation t_i
- ▶ Generate *validation* proof $C_i \models t_i$
 - ▶ Sequence of resolution steps based on linear structure of BDD
- ▶ Initial set of TBDDs

Combine Top-Level BDDs

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Completion

- ▶ When $t_k = \perp$ have proof $C_I \models \perp$

Comparing Proofs

Generated by CDCL Solver

- ▶ Resolution
- ▶ Encode conflict clauses
 - ▶ Increasingly strong constraints on set of satisfying solutions
- ▶ Reach empty clause when detect there is no solution

Generated with BDD-Based Solver

- ▶ Extended resolution
- ▶ Justify each recursive step of BDD algorithm
- ▶ Reach empty clause when reduce formula to BDD leaf \perp

Checking

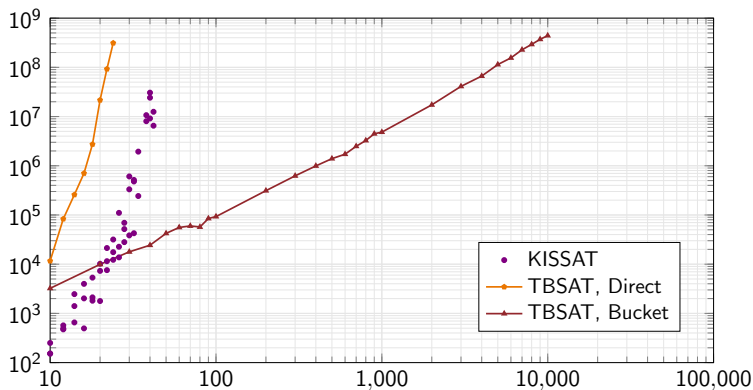
- ▶ Both checked with DRAT/LRAT checkers

TBSAT (Trusted BDD Satisfiability solver)

Implementation

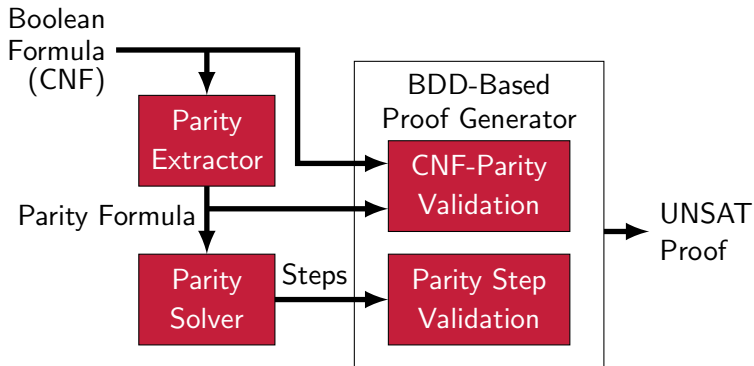
- ▶ TBUDDY: Modified version of BuDDy BDD package
 - ▶ Lind-Nielsen, ca. 1998
- ▶ Support for TBDDs and proof generation
- ▶ C/C++
- ▶ <https://github.com/rebryant/tbuddy-artifact>

Parity Benchmark Proof Complexity



- ▶ Total number of proof steps
 - ▶ Defining clauses + RUP clauses
- ▶ TBSAT with bucket elimination scales polynomially
 - ▶ Checker time \approx Solver time

Integrating Parity Reasoning into Proof-Generating SAT Solver



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

Gaussian Elimination Over GF2

System of Equations $E = \{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_m\}$

$$\mathbf{e}_i : \sum_{j=1,n} a_{i,j} \cdot x_j = b_i$$

Assume

- ▶ $a_{i,j}, x_j \in \{0, 1\}$
- ▶ $a + b \equiv a \oplus b$
- ▶ $a \cdot b \equiv a \wedge b$

Capability

- ▶ Can determine if there are any solutions for x_1, x_2, \dots, x_n

Gaussian Elimination Over GF2

System of Equations $E = \{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_m\}$

$$\mathbf{e}_i : \sum_{j=1,n} a_{i,j} \cdot x_j = b_i$$

Elimination Step

1. Choose pivot equation \mathbf{e}_s and variable x_t such that $a_{s,t} = 1$
2. For each $i \neq s$:

$$\mathbf{e}_i \leftarrow \begin{cases} \mathbf{e}_i & a_{i,t} = 0 \\ \mathbf{e}_s + \mathbf{e}_i, & a_{i,t} = 1 \end{cases}$$

- Guarantees $a_{i,t} = 0$ for all $i \neq s$

3. Remove \mathbf{e}_s from E and repeat until single equation left

Gaussian Elimination Results

Possible Outcomes

1. If encounter degenerate equation
 - ▶ Of form $0 = 1$
 - ▶ Has no solution
2. Otherwise,
 - ▶ Can perform back substitution to find solution

CNF to Parity Constraint Validation

Clauses

- ▶ Suppose clauses $C_{i_1}, C_{i_2}, \dots, C_{i_k}$ encode parity constraint equation \mathbf{e}
- ▶ Have validated BDD representations $t_{i_1}, t_{i_2}, \dots, t_{i_k}$

Form conjunction

$$s = \bigwedge_{1 \leq j \leq k} t_{i_j}$$

- ▶ Also yields proof $C_I \models s$

Represent Constraint

- ▶ Form BDD representation t_j of \mathbf{e}

Validate

- ▶ Generate proof $s \rightarrow t_j$
- ▶ Use to validate term $C_I \models t_j$

Parity Step Validation

Assume

- ▶ Have BDDs t_i and t_j representing equations \mathbf{e}_i and \mathbf{e}_j
- ▶ Satisfying $C_I \models t_i$ and $C_I \models t_j$

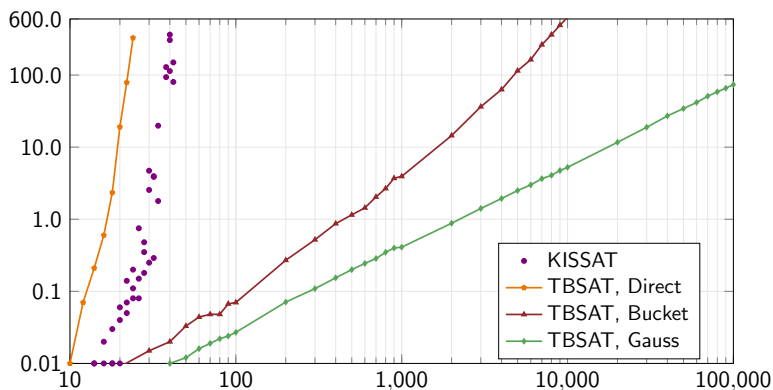
Compute

- ▶ $s \leftarrow \text{ApplyAnd}(t_i, t_j)$
 - ▶ Gives proof $t_i \wedge t_j \rightarrow s$
- ▶ Generate BDD representation t_k of equation $\mathbf{e}_k = \mathbf{e}_i + \mathbf{e}_j$

Validation

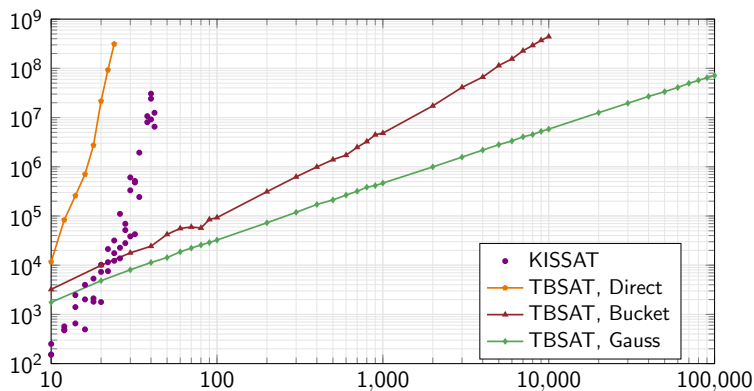
- ▶ Generate proof $s \rightarrow t_k$
- ▶ Combine with other proofs to validate $C_I \models t_k$

Parity Benchmark Runtime



- ▶ $n = 100,000$ in 74 seconds
- ▶ Upper limit: $n = 699,051$
 - ▶ BuDDy limited to $2^{21} - 1$ BDD variables

Parity Benchmark Proof Complexity



- ▶ Total number of proof steps
 - ▶ Defining clauses + RUP clauses
- ▶ Checker time \approx Solver time

Final Thoughts on SAT Solvers

CDCL is the best overall approach

- ▶ Readily generates resolution proofs
- ▶ But, very weak for parity and cardinality constraints

BDDs provide complementary strengths

- ▶ Can generate extended resolution proofs
- ▶ Very strong for parity constraints
- ▶ Some success with cardinality constraints

Future solvers should use combination of methods

- ▶ With unified proof framework
- ▶ Clausal reasoning
- ▶ Constraint reasoning
- ▶ Boolean reasoning

Final Thoughts on Checkable Proofs

Important capability

- ▶ Vital to gain confidence in automated reasoning tools
- ▶ Benefits both tool developers and tool users

SAT community handled this especially well

- ▶ Started with well-established logical framework (resolution)
- ▶ Developed efficient algorithms that integrated well with solvers (RUP)
- ▶ Included more general capabilities (extended resolution)
- ▶ Formulated file formats, tool chain
- ▶ Fostered deployment through competitions

More challenging for other domains

- ▶ Beyond Boolean

Some References

BDDs

- ▶ R. E. Bryant, “[Graph-Based Algorithms for Boolean Function Manipulation](#),” *IEEE Transactions on Computers*, 1986
- ▶ R. E. Bryant, “[Binary Decision Diagrams](#),” *Handbook of Model Checking*, 2018

Proof Generation with BDDs

- ▶ R. E. Bryant and M. J. H. Heule, “[Generating Extended Resolution Proofs with a BDD-Based SAT Solver](#),” *TACAS*, 2021
- ▶ R. E. Bryant, A. Biere, and M. J. H. Heule, “[Clausal Proofs from Pseudo-Boolean Reasoning](#),” *TACAS*, 2022
- ▶ R. E. Bryant, “[TBUDDY: A Proof-Generating BDD Package](#),” *in submission*, 2022