

# Trustworthy Boolean Reasoning

## 2A: Introduction to BDDs

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June, 2022

# 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, ...

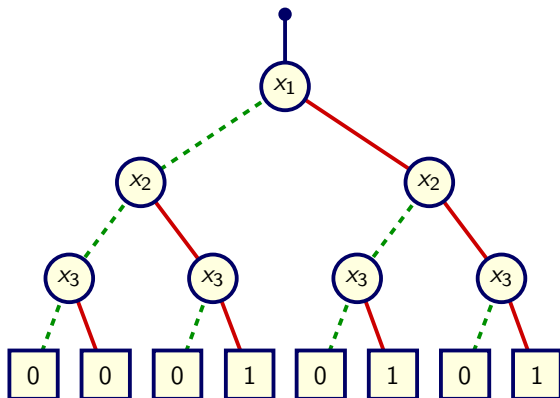
- ▶ Bottom-up approach
  - ▶ Construct canonical representation of problem
  - ▶ Generate solutions
- ▶ Compare to search-based methods
  - ▶ E.g., CDCL
  - ▶ Top-down approaches
  - ▶ Keep branching on variables until find solution

# 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



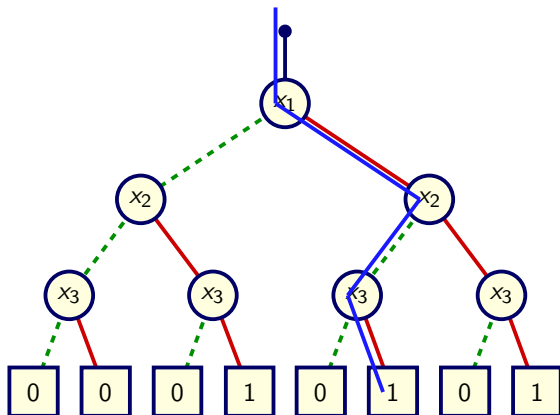
► Size =  $O(2^n)$

# Boolean Function Representations

## Truth Table

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## Decision Tree



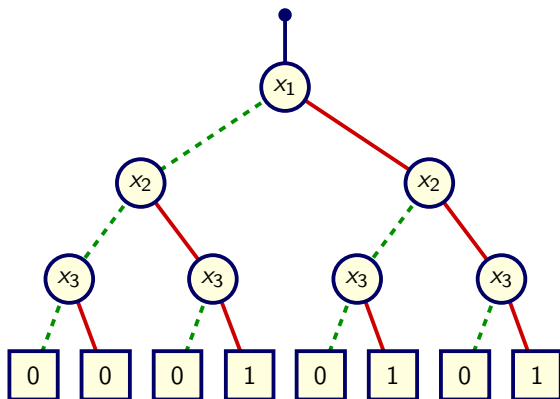
- ▶ Size =  $O(2^n)$
- ▶ Assignment defines path from root to leaf

# Reducing to Canonical Form

Truth Table

$x_1$	$x_2$	$x_3$	$f$
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Graph Representation



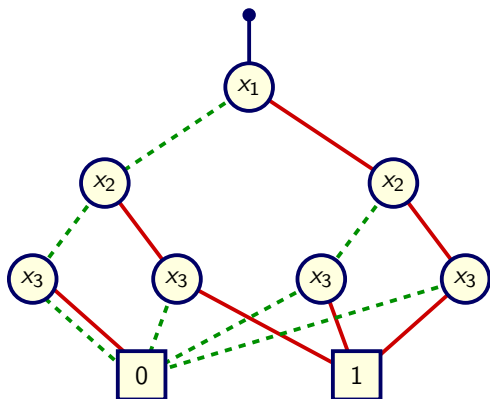
- ▶ Merge isomorphic nodes
- ▶ Eliminate redundant tests

# Reducing to Canonical Form

Truth Table

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Graph Representation



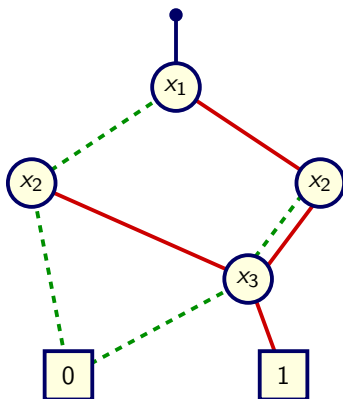
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# Reducing to Canonical Form

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## Graph Representation



- ▶ Merge isomorphic nodes
- ▶ Eliminate redundant tests

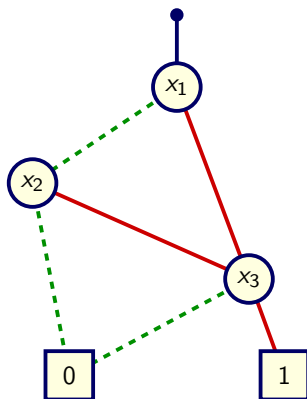


# Reducing to Canonical Form

## Truth Table

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## Graph Representation



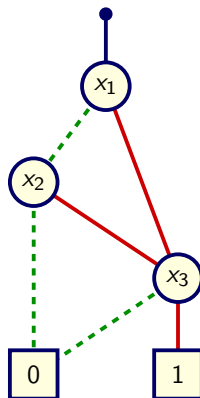
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# Canonical Form

## Truth Table

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## Reduced Ordered Binary Decision Diagram



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

# BDD Representation of Unsatisfiable Formula

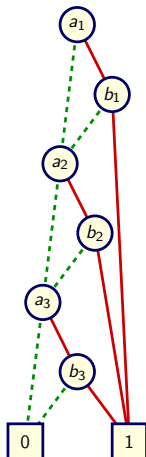
0

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

# Effect of Variable Ordering

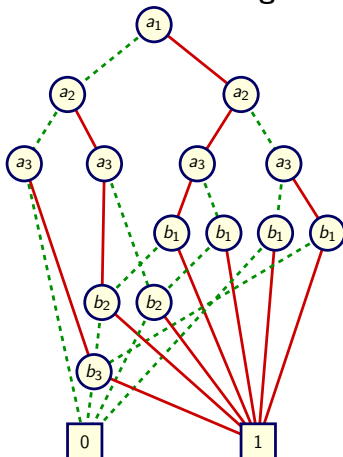
$$(a_1 \wedge b_1) \vee (a_2 \wedge b_2) \vee (a_3 \wedge b_3)$$

**Good Ordering**



► Linear growth

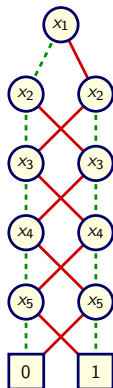
**Bad Ordering**



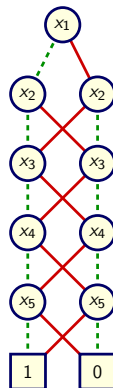
► Exponential growth

# 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

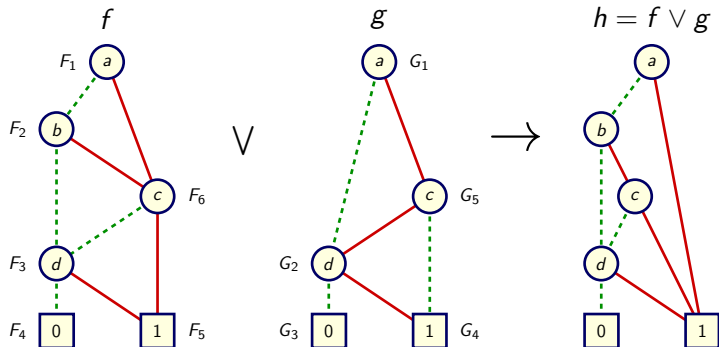
## Key Algorithmic Properties

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

# Apply Algorithm

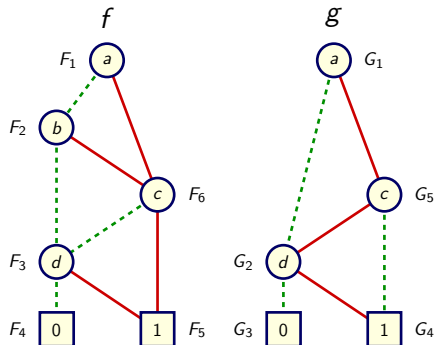
$$h \leftarrow f \odot g$$

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

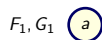


# Apply Algorithm Recursion

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



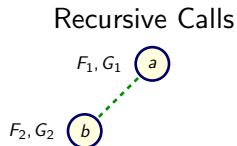
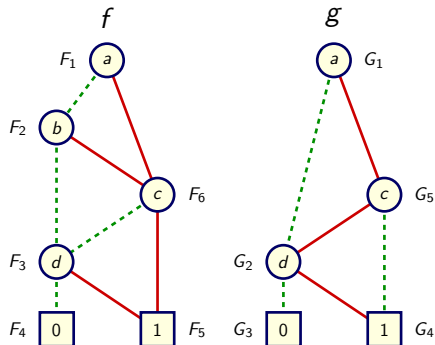
Recursive Calls





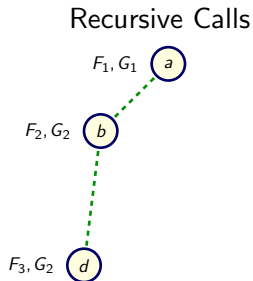
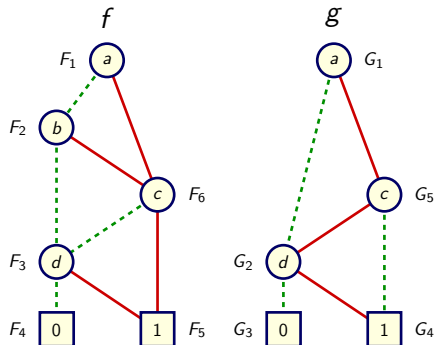
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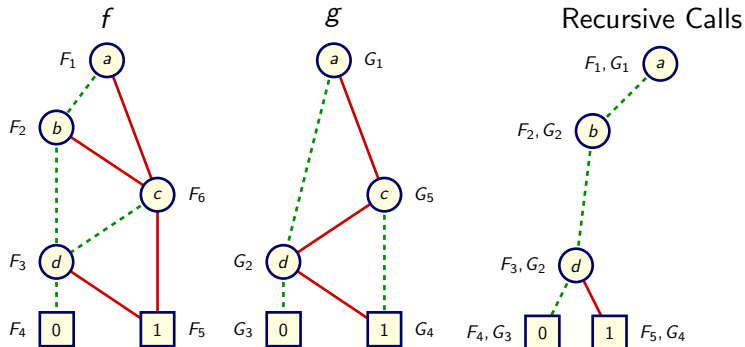
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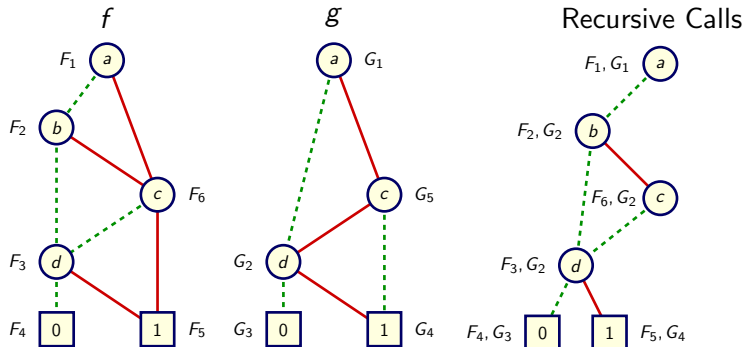
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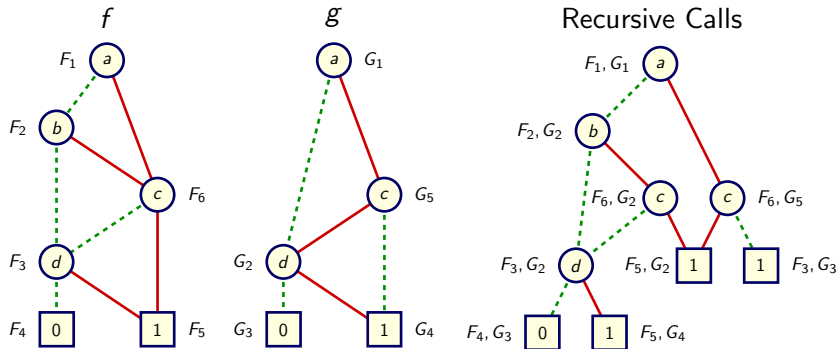
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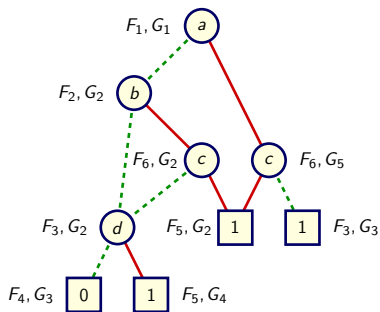
# Apply Algorithm Recursion

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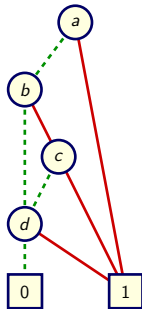


# Apply Algorithm Result

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

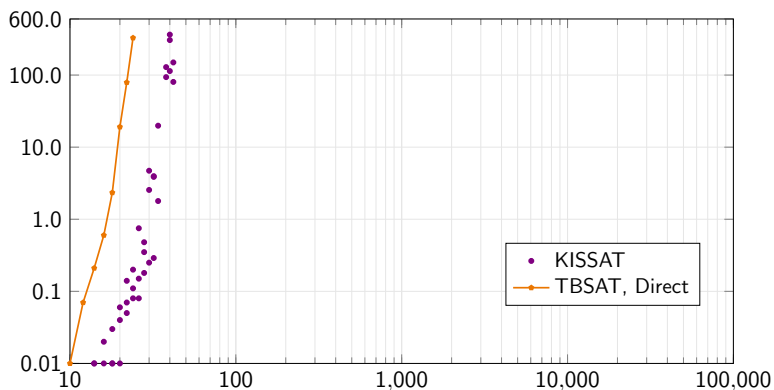
$$s = \bigwedge_{1 \leq i \leq m} t_i$$

- ▶ 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 Evaluation

- ▶ 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  $B_j$

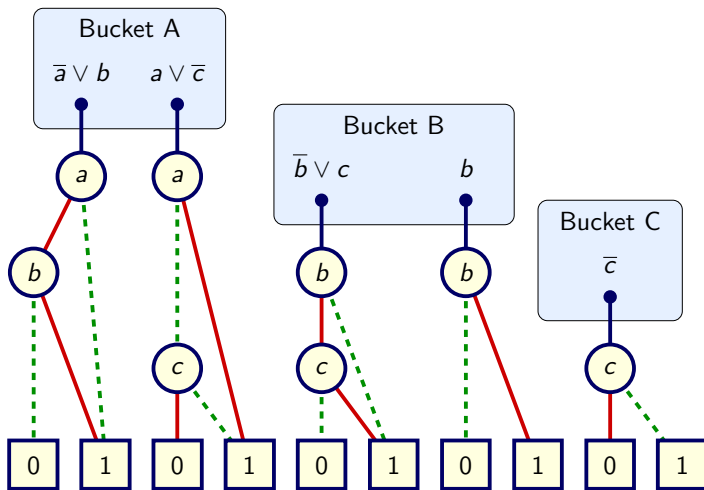
If  $s_j = \perp$  then return UNSAT

Compute  $r_j = \exists x_j s_j$

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

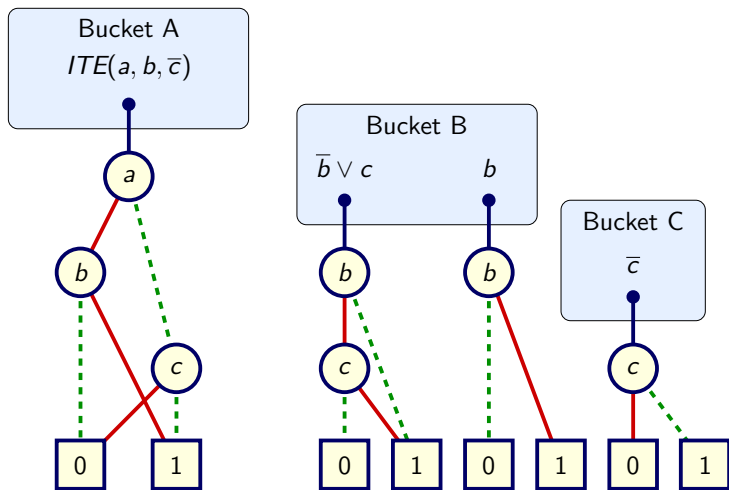
return SAT

# Bucket Evaluation Example



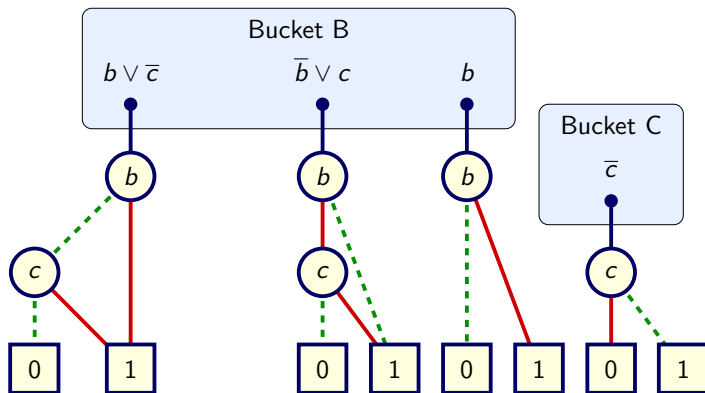
- ▶ Initially: BDD for each input clause
- ▶ In bucket according to root variable

## Bucket Evaluation Example



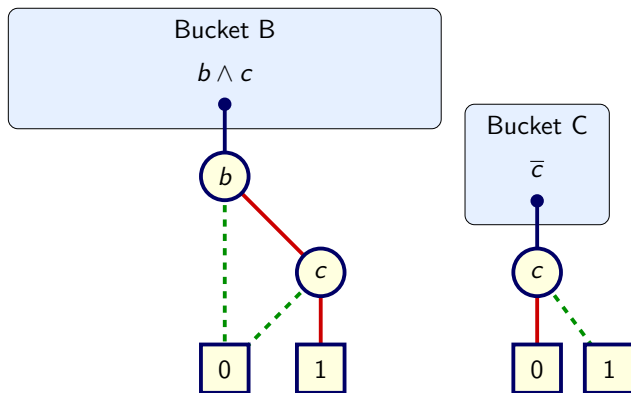
- Conjunct BDDs in topmost bucket A

## Bucket Evaluation Example



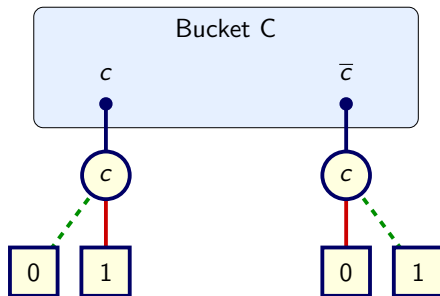
- ▶ Existentially quantify variable  $a$
- ▶ Place result in appropriate bucket

## Bucket Evaluation Example



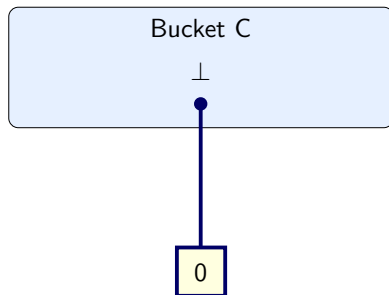
- Conjunct BDDs in bucket  $B$

# Bucket Evaluation Example



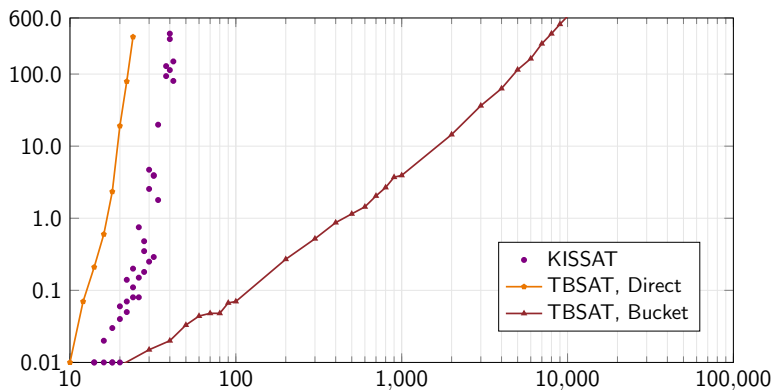
- ▶ Existentially quantify variable  $b$
- ▶ Place result in appropriate bucket

## Bucket Evaluation 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?*