# EECS 219C: Computer-Aided Verification Boolean Satisfiability Solving III \& Binary Decision Diagrams 

Sanjit A. Seshia

EECS, UC Berkeley

Acknowledgments: Lintao Zhang

## DLL Algorithm Pseudo-code



## DLL Algorithm Pseudo-code

```
DLL_iterative()
{
    status = preprocess(); \longleftarrow Pre-processing
    if (status!=UNKNOWN)
        return status;
    while(1) {
        decide_next_branch();\longleftarrow « Branching
        while (true)
        {
            status = deduce();}\longleftarrow<<< Unit propagatio
            if (status == CONFLICT)
            {
                blevel = analyze_conflict();
                if(blevel < 0)
                \mathrm{ else backtrack(blevel);}% & Backtracking
            }
            else if (status == SATISFIABLE)
                return SATISFIABLE;
            else break;
        }
}
```

S. A. Sesh \}

## A Classification of SAT Algorithms

- Davis-Putnam (DP)
- Based on resolution
- Davis-Logemann-Loveland (DLL/DPLL)
- Search-based
- Basis for current most successful solvers
- Stalmarck's algorithm [optional reading]
- More of a "breadth first" search, proprietary algorithm
- Stochastic search
- Local search, hill climbing, etc.
- Unable to prove unsatisfiability (incomplete)


## Proof Generation

- SAT Solvers are complicated pieces of software
- What if there is a bug in the solver? How can we check its output?


## Proof Generation

- If the SAT solver returns "satisfiable", we can check that solution by evaluating the circuit
- If it returns "unsatisfiable", what then?

"You WANT PROOF? I'LL GNE YOU PROOF!"


## Proof

- Starting from facts (clauses), the SAT solver has presumably derived "unsatisfiable" (the empty clause)
- So there must be a way of going step-bystep from input clauses to the empty clause using rules
- In fact, there's only one rule: resolution


# Resolution as a Cut in Implication Graph 



$$
\left(V_{2}+V_{3}{ }^{\prime}+V_{5}^{\prime}+V_{6}\right)
$$

$$
\left(V_{3}^{\prime}+V_{6}+V_{4}\right)
$$

$$
\left(V_{6}+V_{5}^{\prime}+V_{1}^{\prime}\right)
$$

$$
\left(V_{2}+V_{4}^{\prime}+V_{6}+V_{1}\right)
$$

$$
\left(\mathrm{V}_{2}+\mathrm{V}_{4}^{\prime}+\mathrm{V}_{6}+\mathrm{V}_{5}^{\prime}\right)
$$

$$
\left(V_{2}+V_{3}^{\prime}+V_{5}^{\prime}+V_{6}\right)
$$

## Resolution Graph

- Nodes are clauses
- Edges are applications of resolution



## Proof Checker

- Given resolution graph, how to check it?
- Traverse it, checking that each node is correctly obtained from its predecessor nodes using resolution


## Unsatisfiable Core



## Incremental SAT Solving

- Suppose you have not just one SAT problem to solver, but many "slightly differing" problems over the same variables
- Can we re-use the search over many problems?
- i.e. perform only "incremental" work


## Operations Needed

1. Adding clauses
2. Deleting clauses

- Which is easy and which is hard?
- If previous problem is unsat, how does an operation change it?
- If previous is sat?


## Deleting Clauses



## Deleting Clauses



## Engineering Issues

- Too expensive to traverse graph
- Instead, group original clauses into groups
- Each derived clause belongs to all groups that it is resolved from
- Implement with bit-vector


## Binary Decision Diagrams

## Boolean Function Representations

- Syntactic: e.g.: CNF, DNF, Circuit
- Semantic: e.g.: Truth table, Binary Decision Tree, BDD


## Reduced Ordered BDDs

- Invented by Randal E. Bryant in mid-80s
- IEEE Transactions on Computers 1986 paper is one of the most highly cited papers in EECS
- Useful data structure to represent Boolean functions
- Applications in synthesis, verification, program analysis, planning, ...
- Commonly known simply as BDDs
- Many variants of BDDs have proved useful in other tasks
- Links to coding theory (trellises), etc.


## Cofactors

- A Boolean function $F$ of $n$ variables $x_{1}, x_{2}, \ldots$, $\mathrm{X}_{\mathrm{n}}$

$$
F:\{0,1\}^{n} \rightarrow\{0,1\}
$$

- Suppose we define new Boolean functions of $n-1$ variables as follows:

$$
\begin{aligned}
& F_{x_{1}}\left(x_{2}, \ldots, x_{n}\right)=F\left(1, x_{2}, x_{3}, \ldots, x_{n}\right) \\
& F_{x_{1_{1}^{\prime}}^{\prime}}\left(x_{2}, \ldots, x_{n}\right)=F\left(0, x_{2}, x_{3}, \ldots, x_{n}\right)
\end{aligned}
$$

- $F_{x_{1}}$ and $F_{x_{1}}$, are called cofactors of $F$. $F_{x_{1}}$ is the positive cofactor, and $F_{x_{1}}$, is the negative cofactor


## Shannon Expansion

- $F\left(x_{1}, \ldots, x_{n}\right)=x_{i} \cdot F_{x_{i}}+x_{i}^{\prime} \cdot F_{x_{i}^{\prime}}$
- Proof?


## Shannon expansion with many variables

- $F(x, y, z, w)=$
$x y F_{x y}+x^{\prime} y F_{x^{\prime} y}+x y^{\prime} F_{x y^{\prime}}+x^{\prime} y^{\prime} F_{x^{\prime} y^{\prime}}$


## Properties of Cofactors

- Suppose you construct a new function H from two existing functions F and G : e.g.,
$-\mathrm{H}=\mathrm{F}^{\prime}$
$-H=F . G$
$-H=F+G$
- Etc.
- What is the relation between cofactors of H and those of F and G ?


## Very Useful Property

- Cofactor of NOT is NOT of cofactors
- Cofactor of AND is AND of cofactors
- ...
- Works for any binary operator


## BDDs from Truth Tables

Truth Table
Bedinary Decision Tree
Berdered Binary Décision Diagram (OBDD)
(ROBDD, simply called BDD)

## Example: Odd Parity Function



Binary Decision Tree



## Reduction

- Identify Redundancies
- 3 Rules:

1. Merge equivalent leaves
2. Merge isomorphic nodes
3. Eliminate redundant tests

## Merge Equivalent Leaves


" $a$ " is either 0 or 1


Eliminate Redundant Tests



Final ROBDD for Odd Parity Function



## What can BDDs be used for?

- Uniquely representing a Boolean function
- And a Boolean function can represent sets
- Satisfiability solving!


## (RO)BDDs are canonical

- Theorem (R. Bryant): If G, G' are ROBDD's of a Boolean function $f$ with $k$ inputs, using same variable ordering, then $G$ and $\mathrm{G}^{\prime}$ are identical.


## Sensitivity to Ordering

- Given a function with n inputs, one input ordering may require exponential \# vertices in ROBDD, while other may be linear in size.
- Example: $f=x_{1} x_{2}+x_{3} X_{4}+x_{5} x_{6}$
$x_{1}<x_{2}<x_{3}<x_{4}<x_{5}<x_{6}$

$$
x_{1}<x_{4}<x_{5}<x_{2}<x_{3}<x_{6}
$$



## Creating BDDs

- Idea: Build a few core operators and define everything else in terms of those


## Advantage:

- Less programming work
- Easier to add new operators later by writing "wrappers"


## Core Operators

- Just two of them!

1. Restrict(Function F, variable v, constant k)

- Shannon cofactor of F w.r.t. $\mathrm{v}=\mathrm{k}$

2. ITE(Function I, Function T, Function E)

- "if-then-else" operator


## ITE

- Just like:
- "if then else" in a programming language
- A mux in hardware
- ITE(I(x), T(x), E(x))
- If $I(x)$ then $T(x)$ else $E(x)$



## The ITE Function

$$
\begin{aligned}
& \operatorname{ITE}(\mathrm{I}(x), T(x), E(x)) \\
& = \\
& I(x) \cdot T(x)+I^{\prime}(x) \cdot E(x)
\end{aligned}
$$

## What good is the ITE?

- How do we express
- NOT?
- OR?
- AND?


## How do we implement ITE?

- Divide and conquer!
- Use Shannon cofactoring...
- Recall: Operator of cofactors is Cofactor of operators...


## ITE Algorithm

ITE (bdd I, bdd T, bdd E) \{
if (terminal case) \{ return computed result;
\}
else \{ // general case
Let $x$ be the topmost variable of I, T, E;
PosFactor $=\operatorname{ITE}\left(I_{x}, T_{x}, E_{x}\right)$;
NegFactor $=\operatorname{ITE}\left(\mathrm{I}_{\mathrm{x}^{\prime}}, \mathrm{T}_{\mathrm{x}^{\prime}}, \mathrm{E}_{\mathrm{x}^{\prime}}\right)$;
$R$ = new node labeled by $x$;
R.low = NegFactor; // set 0-child
R.high = PosFactor; // set 1-child

Reduce(R); // apply reduction rules return R ;

## Terminal Cases

- $\operatorname{ITE}(1, \mathrm{~T}, \mathrm{E})=$
- ITE(0, T, E) =
- ITE $(\mathrm{I}, \mathrm{T}, \mathrm{T})=$
- ITE(I, 1, 0) =
- ...


## General Case

- Still need to do cofactor (Restrict)
- Easy operation in the case of ITE
- We are cofactoring w.r.t. the top variable only
- Cofactoring w.r.t. an internal variable is trickier


## Practical Issues

- Previous calls to ITE are cached
- "memoization"
- Every BDD node created goes into a "unique table"
- Before creating a new node R, look up this table
- Avoids need for reduction


## Sharing: Multi-Rooted DAG

- BDD for 4-bit adder
- Each output bit (of the sum \& carry) is a distinct rooted BDD
- But they share subDAGs



## More on BDDs

- Circuit width and bounds on BDD size
- Dynamically changing variable ordering
- Some BDD variants


## Bounds on BDD Size: Warm-up

- Suppose the number of nodes at any level in a BDD is bounded above by B
- Then, what is an upper bound on the total number of nodes in the BDD?


## Cross-section of a BDD at level i

- Suppose a BDD represents Boolean function $\mathrm{F}\left(\mathrm{x}_{1}, \mathrm{x}_{2}, \ldots, \mathrm{x}_{\mathrm{n}}\right)$ with variable order $x_{1}<x_{2}<\ldots<x_{n}$
- Size of cross section of the BDD at level i is the number of distinct Boolean functions $\mathrm{F}^{\prime}$ that depend on $\mathrm{x}_{\mathrm{i}}$ given by
$F^{\prime}\left(x_{i}, x_{i+1}, \ldots, x_{n}\right)=F\left(v_{1}, v_{2}, \ldots, v_{i-1}, x_{i}, \ldots, x_{n}\right)$
for some Boolean constants $v_{i}$ 's (in $\{0,1\}$ )


## Circuit Width

- Consider a circuit representation of a Boolean function F
- Impose a linear order on the gates of the circuit
- Primary inputs and outputs are also considered as "gates" and primary output is at the end of the ordering
- Forward cross section at a gate g: set of wires going from output of $g_{1}$ to input of $g_{2}$ where $g_{1} \leq g<g_{2}$
- Similarly define reverse cross section: set of wires going from output of $g_{1}$ to input of $g_{2}$ where $g_{2} \leq g<g_{1}$
- Forward width $\left(w_{f}\right)$ : maximum forward cross section size
- Similarly, reverse width $\mathrm{w}_{\mathrm{r}}$


## BDD Upper Bounds from Circuit Widths

- Theorem: Let a circuit representing F with $n$ variables have forward width $w_{f}$ and reverse width $w_{r}$ for some linear order $L$ on its gates. Then, there is a BDD representing F of size bounded above by $n .2^{w_{f} \cdot 2^{w r}}$


## BDD Ordering in Practice

- If we can derive a small upper bound using circuit width, then that's fine
- Use the corresponding linear order on the variables
- What if we can't?
- There are many BDD variable ordering heuristics around, but the most common way to deal with variable ordering is to start with something "reasonable" and then swap variables around to improve BDD size
- DYNAMIC VARIABLE REORDERING $\rightarrow$ SIFTING


## Sifting

- Dynamic variable re-ordering, proposed by R. Rudell
- Based on a primitive "swap" operation that interchanges $x_{i}$ and $x_{i+1}$ in the variable order
- Key point: the swap is a local operation involving only levels i and i+1
- Overall idea: pick a variable $x_{i}$ and move it up and down the order using swaps until the process no longer improves the size
- A "hill climbing" strategy


## Some BDD Variants

- Free BDDs (FBDDs)
- Relax the restriction that variables have to appear in the same order along all paths
- How can this help?
- Is it canonical?


## Some BDD Variants

- MTBDD (Multi-Terminal BDD)
- Represents function of Boolean variables with non-Boolean value (integer, rational)
- E.g., input-dependent delay in a circuit, transition probabilities in a Markov chain
- Similar reduction / construction rules to BDDs


## Some BDD packages

- CUDD - from Colorado University, Fabio Somenzi's group
- BuDDy - from IT Univ. of Copenhagen

