

# Combinational Equivalence Checking

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EE 709: Testing & Verification of VLSI Circuits

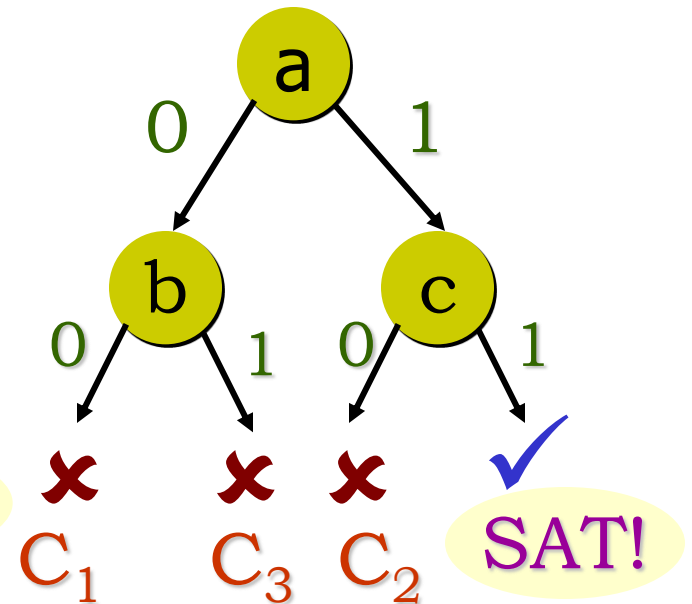
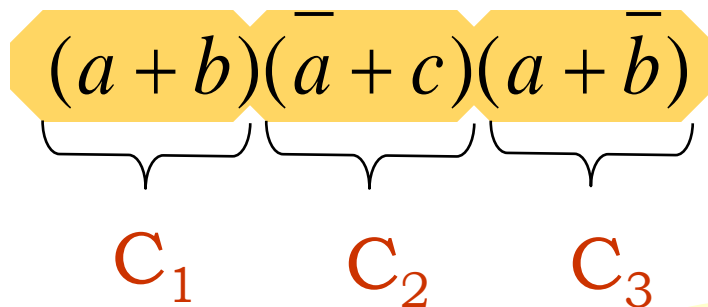
Lecture – 10 (Jan 24, 2012)

# DPLL algorithm for SAT

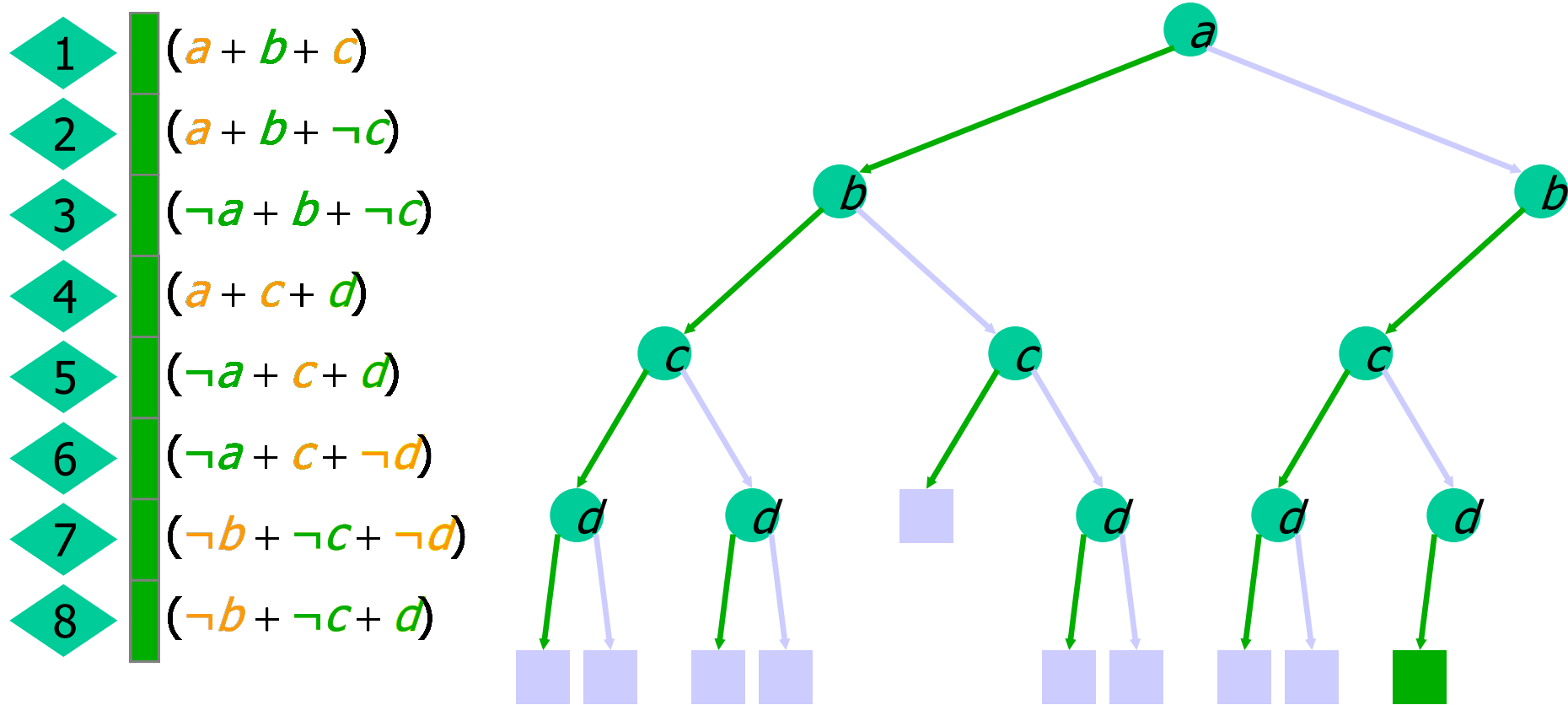
[Davis, Putnam, Logemann, Loveland 1960,62]

*Given* : CNF formula  $f(v_1, v_2, \dots, v_k)$  , and an ordering function *Next\_Variable*

Example :

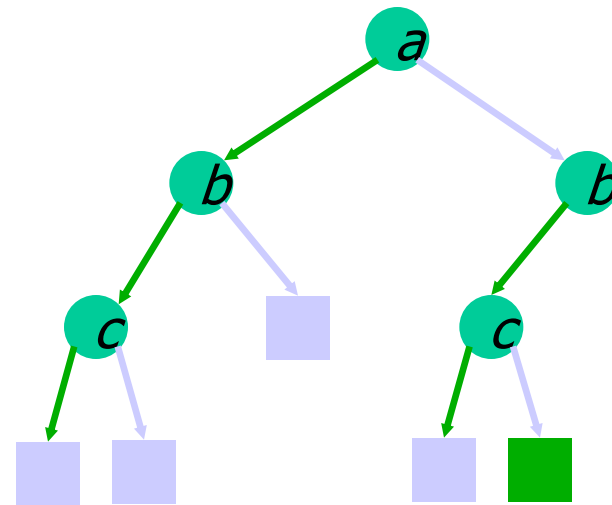


# Basic Backtracking Search



# Basic Search with Implications

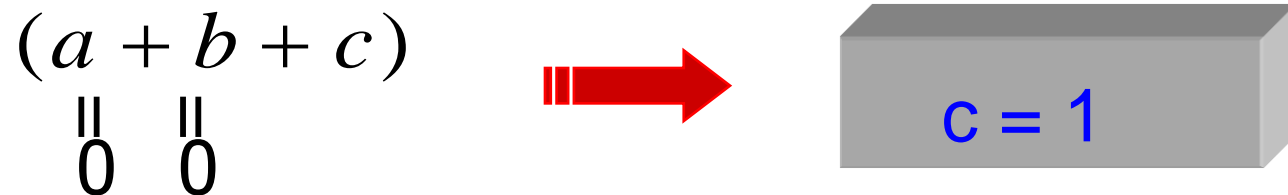
- 1  $(a + b + c)$
- 2  $(a + b + \neg c)$
- 3  $(\neg a + b + \neg c)$
- 4  $(a + c + d)$
- 5  $(\neg a + c + d)$
- 6  $(\neg a + c + \neg d)$
- 7  $(\neg b + \neg c + \neg d)$
- 8  $(\neg b + \neg c + d)$



# DPLL algorithm: Unit clause rule

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*Rule:* Assign to *true* any single literal clauses.



Apply Iteratively: *Boolean Constraint Propagation (BCP)*

$$a(\bar{a} + c)(\bar{b} + c)(a + b + \bar{c})(\bar{c} + e)(\bar{d} + e)(c + d + \bar{e})$$

$$\downarrow$$
$$c(\bar{b} + c)(\bar{c} + e)(\bar{d} + e)(c + d + \bar{e})$$

$$\downarrow$$
$$e(\bar{d} + e)$$

# Pure Literal Rule

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- A variable is *pure* if its literals are either all positive or all negative
- Satisfiability of a formula is unaffected by assigning pure variables the values that satisfy all the clauses containing them


$$\varphi = (a + c)(b + c)(b + \neg d)(\neg a + \neg b + d)$$

- Set  $c$  to 1; if  $\varphi$  becomes unsatisfiable, then it is also unsatisfiable when  $c$  is set to 0.

# Resolution (original DP)

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- Iteratively apply resolution (consensus) to **eliminate one variable each time**
  - i.e., consensus between all pairs of clauses containing  $x$  and  $\neg x$
  - formula satisfiability is **preserved**
- Stop applying resolution when,
  - Either empty clause is derived  $\Rightarrow$  instance is **unsatisfiable**
  - Or only clauses satisfied or with pure literals are obtained  $\Rightarrow$  instance is **satisfiable**

$$\varphi = (a + c)(b + c)(d + c)(\neg a + \neg b + \neg c)$$


Eliminate variable  $c$

$$\begin{aligned}\varphi_1 &= (a + \neg a + \neg b)(b + \neg a + \neg b)(d + \neg a + \neg b) \\ &= (d + \neg a + \neg b)\end{aligned}$$

Instance is **SAT !**

# Stallmarck's Method (SM) in CNF

- Recursive application of the **branch-merge rule** to each variable with the goal of identifying **common conclusions**

$$\varphi = (a + b)(\neg a + c)(\neg b + d)(\neg c + d)$$

Try  $a = 0$ :  $(a = 0) \Rightarrow (b = 1) \Rightarrow (d = 1)$        $C(a = 0) = \{a = 0, b = 1, d = 1\}$

Try  $a = 1$ :  $(a = 1) \Rightarrow (c = 1) \Rightarrow (d = 1)$        $C(a = 1) = \{a = 1, c = 1, d = 1\}$

$$C(a = 0) \cap C(a = 1) = \{d = 1\}$$

Any assignment to variable  $a$  implies  $d = 1$ .  
Hence,  $d = 1$  is a **necessary** assignment !

Recursion can be of arbitrary depth



# Recursive Learning (RL) in CNF

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- Recursive evaluation of **clause satisfiability** requirements for identifying **common assignments**

$$\varphi = (a + b)(\neg a + d)(\neg b + d)$$

Try  $a = 1$ :             $(a = 1) \Rightarrow (d = 1)$              $C(a = 1) = \{a = 1, d = 1\}$

Try  $b = 1$ :             $(b = 1) \Rightarrow (d = 1)$              $C(b = 1) = \{b = 1, d = 1\}$

$C(a = 1) \cap C(b = 1) = \{d = 1\}$       Every way of satisfying  $(a + b)$  implies  $d = 1$ .  
Hence,  $d = 1$  is a **necessary** assignment !

**Recursion can be of arbitrary depth**

# SM vs. RL

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- Both complete procedures for SAT
- Stallmarck's method:
  - hypothetical reasoning based on variables
- Recursive learning:
  - hypothetical reasoning based on clauses
- Both can be integrated into backtrack search algorithms

# Local Search

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- Repeat  $M$  times:
  - Randomly pick complete assignment
  - Repeat  $K$  times (and while exist unsatisfied clauses):
    - Flip variable that will satisfy largest number of unsat clauses

$$\varphi = (a + b)(\neg a + c)(\neg b + d)(\neg c + d)$$

Pick random assignment

$$\varphi = (a + b)(\neg a + c)(\neg b + d)(\neg c + d)$$

Flip assignment on  $d$

$$\varphi = (a + b)(\neg a + c)(\neg b + d)(\neg c + d)$$

Instance is **satisfied** !

# Comparison

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- Local search is **incomplete**
  - If instances are known to be SAT, local search can be competitive
- Resolution is in general **impractical**
- Stallmarck's Method (SM) and Recursive Learning (RL) are in general **slow**, though **robust**
  - SM and RL can derive too much **unnecessary** information
- For most EDA applications **backtrack search (DP)** is currently the most promising approach !
  - **Augmented with techniques for inferring new clauses/implicates (i.e. learning) !**

# Techniques for Backtrack Search

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- Conflict analysis
  - Clause/implicate recording
  - Non-chronological backtracking
- Incorporate and **extend** ideas from:
  - Resolution
  - Recursive learning
  - Stallmarck's method
- Formula simplification & Clause inference [Li,AAAI00]
- Randomization & Restarts [Gomes&Selman,AAAI98]

# Clause Recording

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- During backtrack search, for each conflict **create clause that explains and prevents recurrence of same conflict**

$$\varphi = (a + b)(\neg b + c + d)(\neg b + e)(\neg d + \neg e + f)\dots$$

Assume (decisions)  $c = 0$  and  $f = 0$

Assign  $a = 0$  and imply assignments

A conflict is reached:  $(\neg d + \neg e + f)$  is **unsat**

$$(a = 0) \wedge (c = 0) \wedge (f = 0) \Rightarrow (\varphi = 0)$$

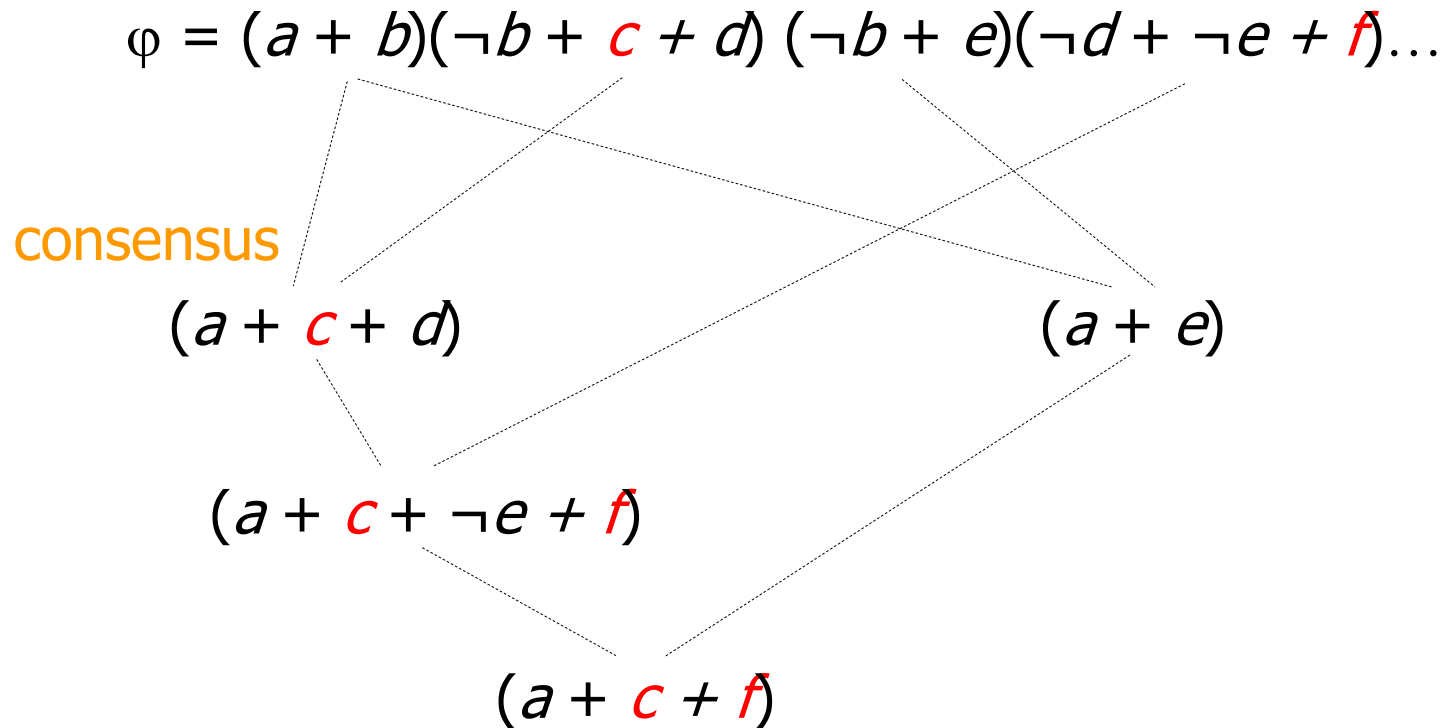
$$(\varphi = 1) \Rightarrow (a = 1) \vee (c = 1) \vee (f = 1)$$

**$\therefore$  create new clause:  $(a + c + f)$**



# Clause Recording

- Clauses derived from conflicts can also be viewed as the result of applying **selective consensus**



# Non-Chronological Backtracking

- During backtrack search, in the presence of conflicts, backtrack to one of the causes of the conflict

$$\varphi = (a + b)(\neg b + c + d)(\neg b + e)(\neg d + \neg e + f) \\ (a + c + f)(\neg a + g)(\neg g + b)(\neg h + j)(\neg i + k) \dots$$

Assume (decisions)  $c = 0, f = 0, h = 0$  and  $i = 0$

Assignment  $a = 0$  caused conflict  $\Rightarrow$  clause  $(a + c + f)$  created  
 $(a + c + f)$  implies  $a = 1$

A conflict is again reached:  $(\neg d + \neg e + f)$  is unsat

$$(a = 1) \wedge (c = 0) \wedge (f = 0) \Rightarrow (\varphi = 0)$$

$$(\varphi = 1) \Rightarrow (a = 0) \vee (c = 1) \vee (f = 1)$$

$\therefore$  create new clause:  $(\neg a + c + f)$



# Non-Chronological Backtracking

Created clauses:  $(a + c + \bar{f})$  and  $(\neg a + c + \bar{f})$

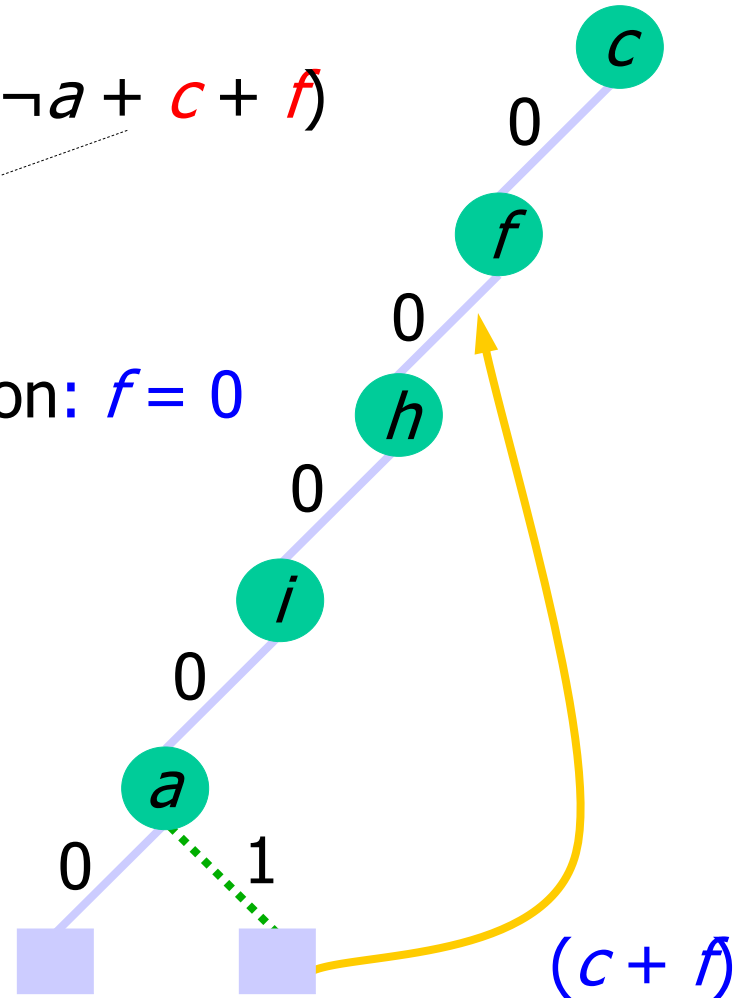
Apply consensus:

new **unsat** clause  $(c + \bar{f})$

$\therefore$  backtrack to most recent decision:  $f = 0$

$\therefore$  created clauses/implicates:

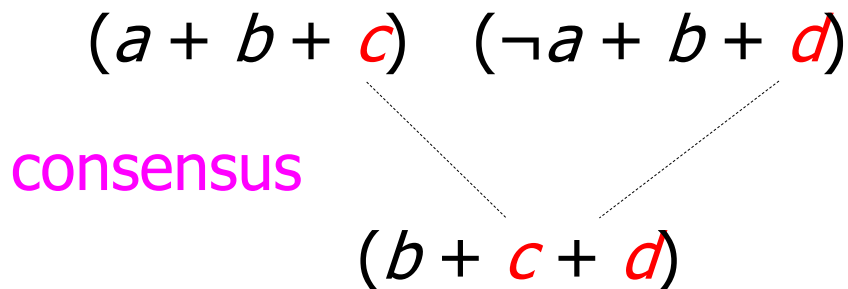
$(a + c + \bar{f})$ ,  
 $(\neg a + c + \bar{f})$ , and  
 $(c + \bar{f})$



# Ideas from other Approaches

- Resolution, Stallmarck's method and recursive learning can be incorporated into **backtrack search (DP)**
  - create additional clauses/implicates
    - anticipate and prevent conflicting conditions
    - identify necessary assignments
    - allow for non-chronological backtracking

## Resolution within DP:



**$(b + c + d)$  Unit clause !**

Clause provides explanation for necessary assignment  $b = 1$

# Stallmarck's Method within DP

$$\varphi = (a + b + e)(\neg a + c + f)(\neg b + d)(\neg c + d + g)$$

Implications:

$$(a = 0) \wedge (e = 0) \Rightarrow (b = 1) \Rightarrow (d = 1)$$

$$(a = 1) \wedge (f = 0) \Rightarrow (c = 1)$$
$$(c = 1) \wedge (g = 0) \Rightarrow (d = 1)$$

$$(e = 0) \wedge (f = 0) \wedge (g = 0) \Rightarrow (d = 1)$$

Clausal form:

**$(e + f + g + d)$  Unit clause !**

Clause provides explanation  
for necessary assignment  $d = 1$

consensus

$$(b + e + c + f)$$

$$(d + e + c + f)$$

$$(e + f + g + d)$$

# Recursive Learning within DP

$$\varphi = (a + b + c)(\neg a + d + e)(\neg b + d + c)$$

Implications:

$$(a = 1) \wedge (e = 0) \Rightarrow (d = 1)$$

$$(b = 1) \wedge (c = 0) \Rightarrow (d = 1)$$

$$(c = 0) \wedge ((e = 0) \wedge (c = 0)) \Rightarrow (d = 1)$$

consensus

$$(b + c + e + d)$$

consensus

$$(c + e + d)$$

Clausal form:

$(c + e + d)$  Unit clause !

Clause provides explanation  
for necessary assignment  $d = 1$

# The Power of Consensus

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- Most search pruning techniques can be explained as particular ways of applying selective consensus
  - Conflict-based clause recording
  - Non-chronological backtracking
  - Extending Stallmarck's method to backtrack search
  - Extending recursive learning to backtrack search
  - Clause inference conditions
- General consensus is computationally too expensive !
- Most techniques indirectly identify which consensus operations to apply !
  - To create new clauses/implicates
    - To identify necessary assignments

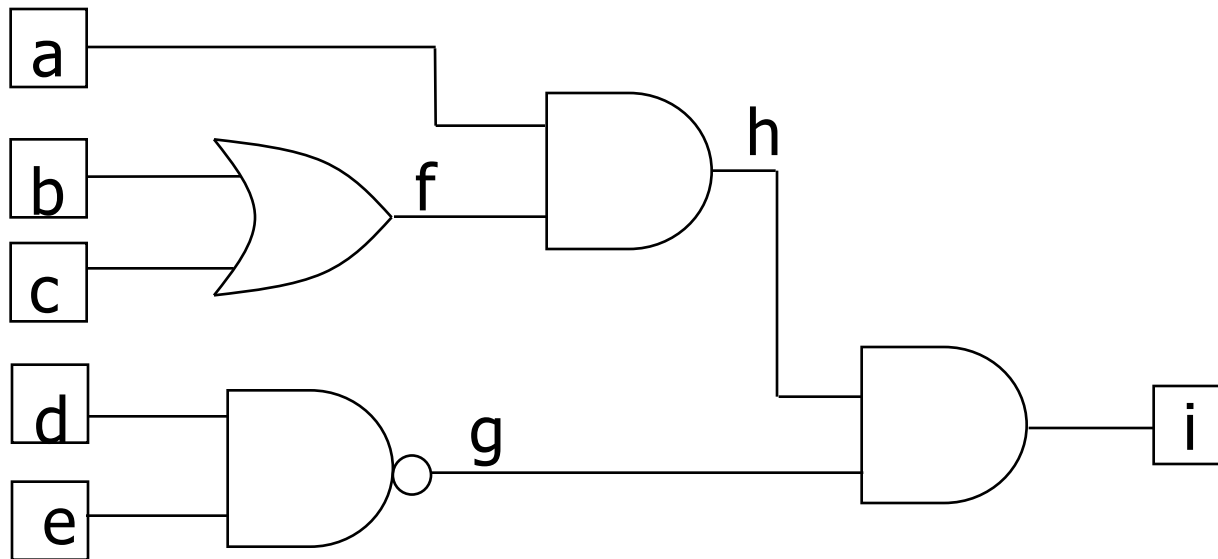
# SAT Solvers Today

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- Capacity:
  - Formulas upto a *million variables* and *3-4 million clauses* can be solved in *few hours*
  - Only for *structured instances* e.g. derived from real-world circuits & systems
- Tool offerings:
  - ◆ Public domain
    - GRASP : Univ. of Michigan
    - SATO: Univ. of Iowa
    - zChaff: Princeton University
    - BerkMin: Cadence Berkeley Labs.
  - ◆ Commercial
    - PROVER: Prover Technologies

# Solving circuit problems as SAT

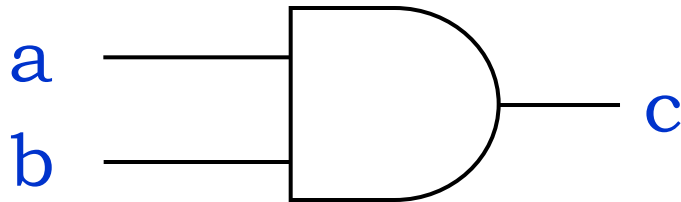
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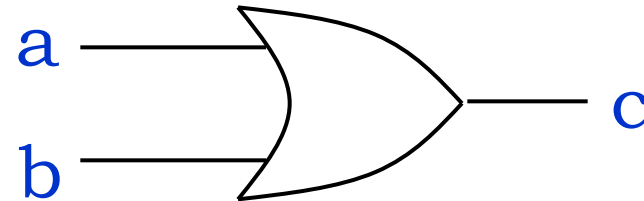
Input Vector Assignment ? ➔ Primary Output 'i' to 1 ?

# SAT formulas for simple gates

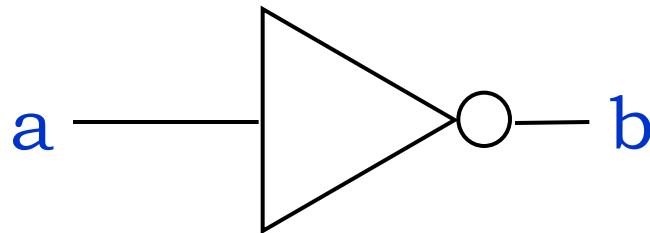
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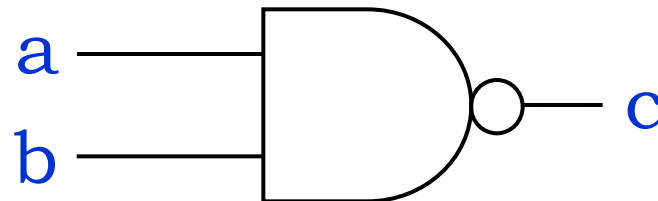
$$(\bar{c} + a)(\bar{c} + b)(c + \bar{a} + \bar{b})$$



$$(c + \bar{a})(c + \bar{b})(\bar{c} + a + b)$$



$$(a + b)(\bar{a} + \bar{b})$$

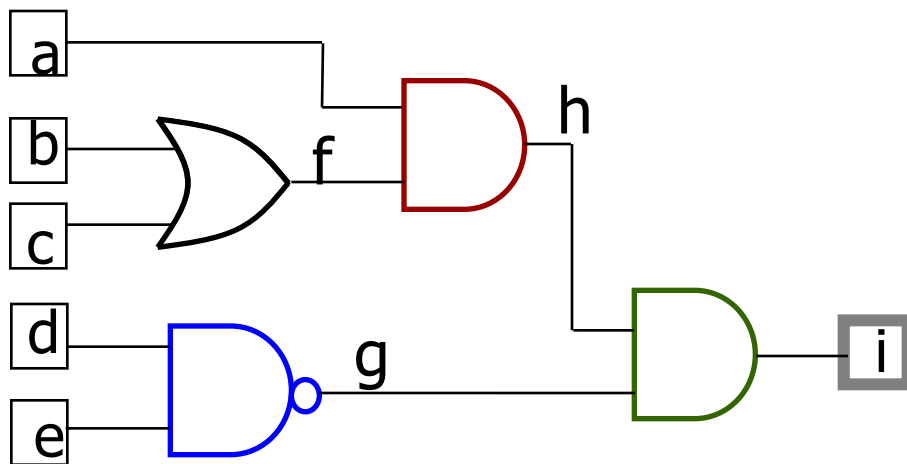


$$(c + a)(c + b)(\bar{c} + \bar{a} + \bar{b})$$



# Solving circuit problems as SAT

- Set of clauses representing function of each gate
- Unit literal clause asserting output to '1'



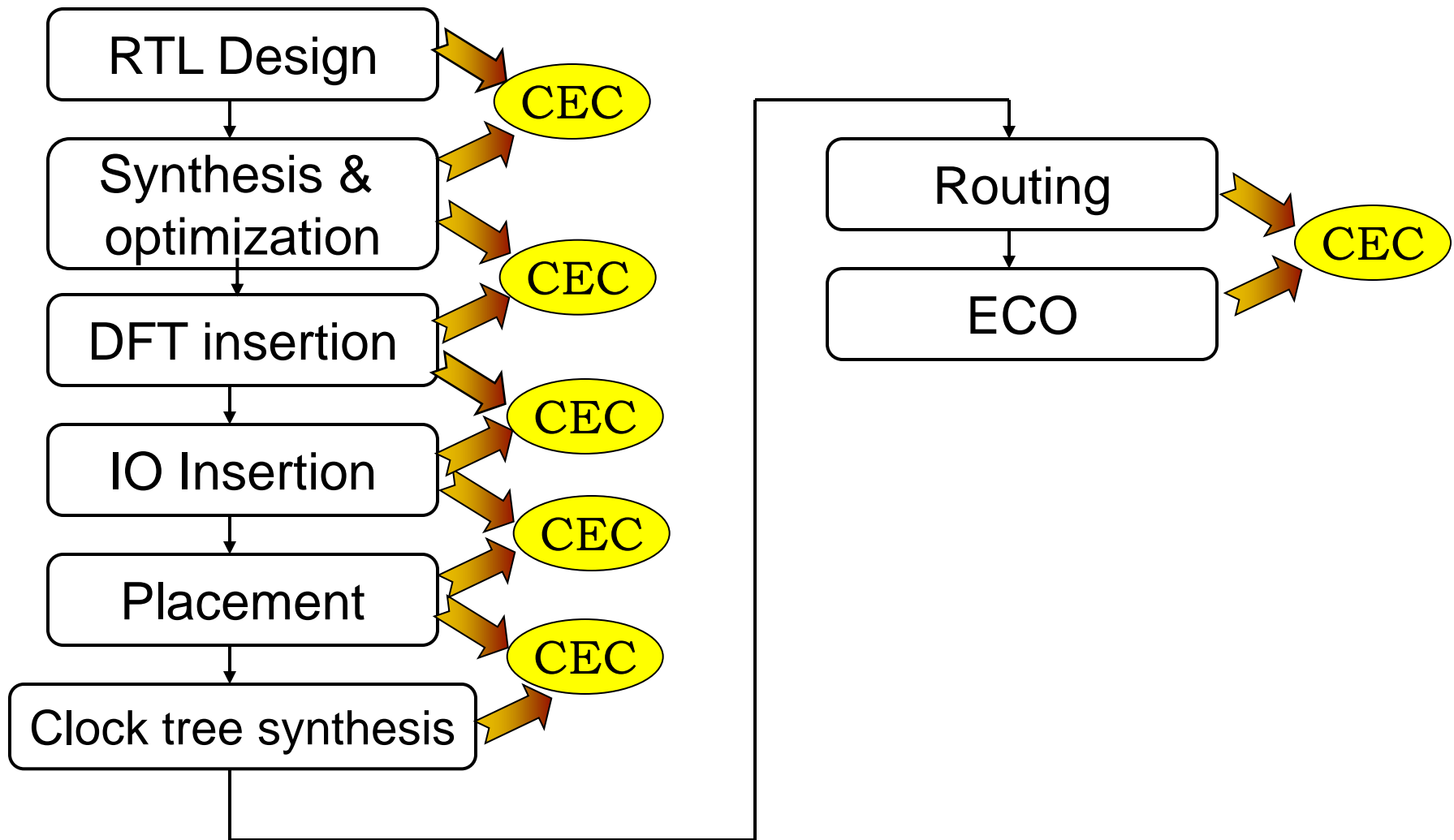
$$\begin{aligned} &(\bar{b} + f)(\bar{c} + f)(b + c + \bar{f}) \\ &(d + g)(e + g)(\bar{d} + \bar{e} + \bar{g}) \\ &(a + \bar{h})(f + \bar{h})(\bar{a} + \bar{f} + h) \\ &(h + \bar{i})(g + \bar{i})(\bar{h} + \bar{g} + i) \\ &(i) \end{aligned}$$

# Combinational Equivalence Checking (CEC)

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- Currently most practical and pervasive equivalence checking technology
- Nearly full automation possible
- Designs of up to several million gates verified in a few hours or minutes
- Hierarchical verification deployed
- Full chip verification possible
- **Key methodology:** Convert sequential equivalence checking to a CEC problem!
  - Match Latches & extract comb. portions for EC

# CEC in Today's ASIC Design Flow



# Major Industrial Offerings of CEC

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- Formality (*Synopsys*)
- Conformal Suite (*Verplex, now Cadence*)
- FormalPro (*Mentor Graphics*)
- Typical capabilities of these tools:
  - Can handle circuits of up to **several million gates flat** in up to a few hours of runtime
  - Comprehensive **debug tool** to pinpoint error-sources
  - **Counter-example display & cross-link** of RTL and gate-level netlists for easier debugging
  - Ability to **checkpoint** verification process and restart from same point later
  - **What if** capability (unique to FormalPro)

# Combinational Equivalence Checking

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- Functional Approach
  - transform output functions of combinational circuits into a unique (**canonical**) representation
  - two circuits are equivalent if their representations are identical
  - efficient canonical representation: BDD
- Structural
  - identify **structurally similar** internal points
  - prove internal points (cut-points) *equivalent*
  - find implications

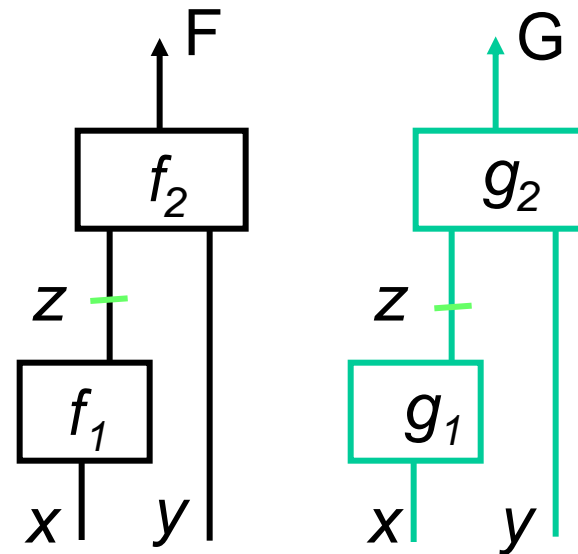
# Functional Equivalence

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- If BDD can be constructed for each circuit
  - represent each circuit as *shared* (multi-output) BDD
    - ❖ **use the same variable ordering !**
  - BDDs of both circuits must be *identical*
- If BDDs are too large
  - cannot construct BDD, memory problem
  - use partitioned BDD method
    - decompose circuit into smaller pieces, each as BDD
    - check equivalence of internal points

# Functional Decomposition

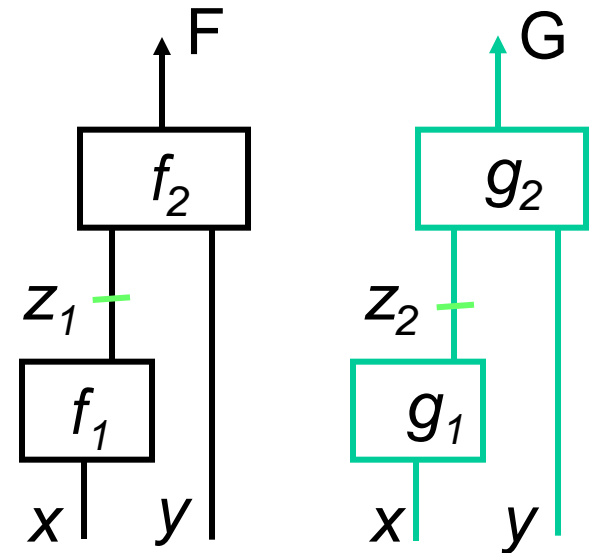
- Decompose each function into *functional* blocks
  - represent each block as a BDD (*partitioned BDD* method)
  - **define cut-points** ( $z$ )
  - verify equivalence of blocks at cut-points
  - starting at primary inputs



# Cut-Points Resolution Problem

- If *all pairs* of cut-points  $(z_1, z_2)$  are equivalent
  - so are the two functions,  $F, G$
- If *intermediate* functions  $(f_2, g_2)$  are not equivalent
  - the functions  $(F, G)$  may still be equivalent
  - this is called **false negative**

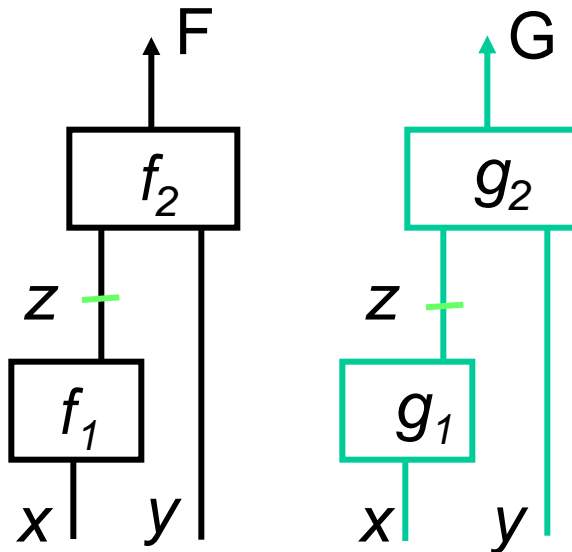
- Why do we have false negative ?
  - functions are represented in terms of *intermediate* variables
  - to prove/disprove equivalence must represent the functions in terms of *primary inputs* (BDD composition)





# Cut-Point Resolution – Theory

- Let  $f_1(x)=g_1(x) \quad \forall x$ 
  - if  $f_2(z,y) \equiv g_2(z,y), \quad \forall z,y$  then  $f_2(f_1(x),y) \equiv g_2(f_1(x),y) \Rightarrow F \equiv G$
  - if  $f_2(z,y) \neq g_2(z,y), \quad \forall z,y \quad \not\Rightarrow \quad f_2(f_1(x),y) \neq g_2(f_1(x),y) \not\Rightarrow F \neq G$



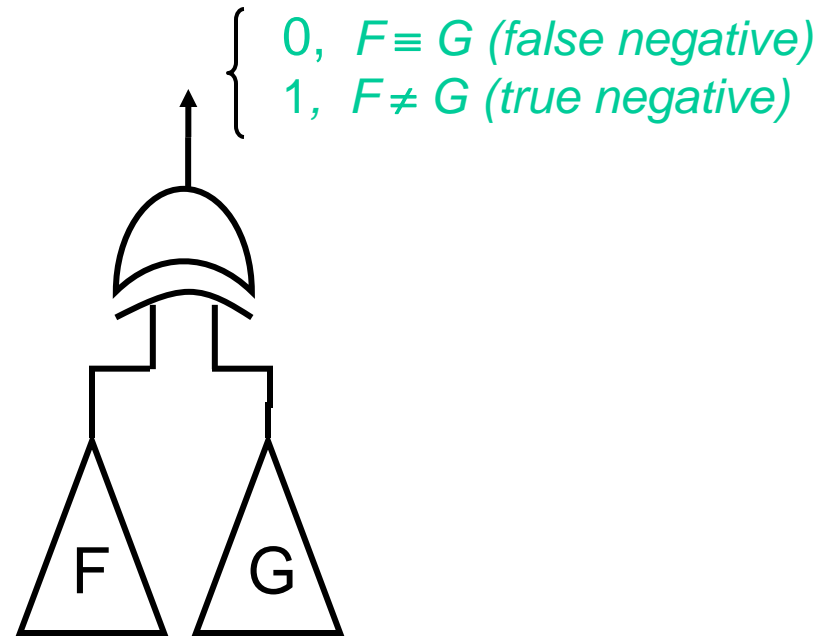
We cannot say if  $F \equiv G$  or not

- *False negative*
  - two functions are equivalent, but the verification algorithm declares them as different.

# Cut-Point Resolution

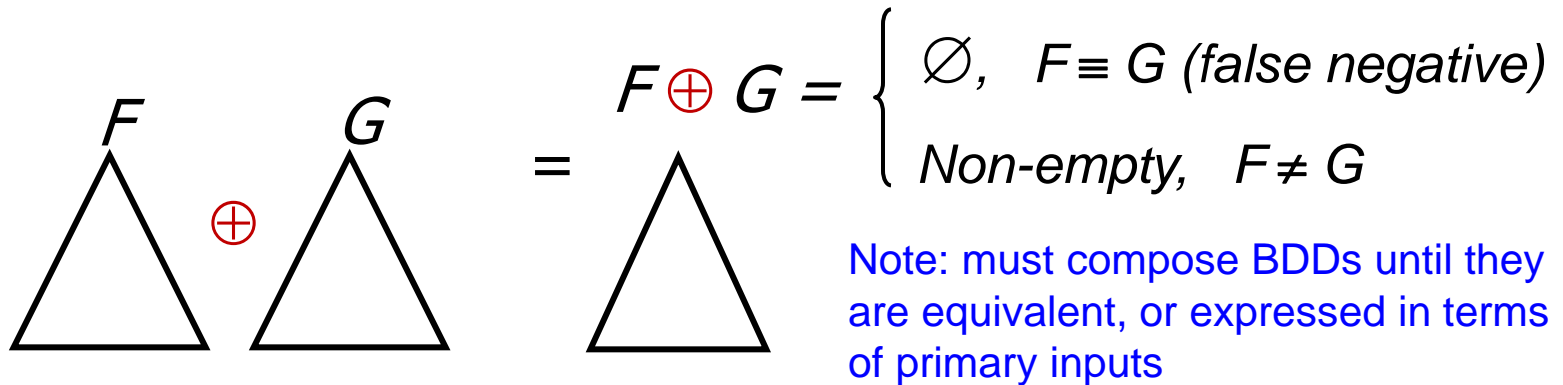
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- How to verify if negative is *false* or *true* ?
- **Procedure 1:** create a miter (XOR) between two potentially equivalent nodes/functions
  - perform ATPG test for *stuck-at 0*
  - find test pattern to prove  $F \neq G$
  - efficient for true negative
  - (gives *test vector*, a proof)
  - inefficient when there is no test



# Cut-Point Resolution

- Procedure 2: **create a BDD for  $F \oplus G$** 
  - perform satisfiability analysis (SAT) of the BDD
    - if BDD for  $F \oplus G = \emptyset$ , problem is *not* satisfiable, *false negative*
    - BDD for  $F \oplus G \neq \emptyset$ , problem is satisfiable, *true negative*



- the SAT solution, if exists, provides a *test vector* (proof of non-equivalence) – as in ATPG
- unlike the ATPG technique, it is effective for false negative (the BDD is empty!)

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# Thank you

