

Quantified Boolean Formula: Evaluation, Certification, and Applications

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Outline

□ Satisfiability (SAT)

- Conjunctive Normal Form (CNF)
- SAT solving and Craig interpolation
- Application
 - Functional dependency

□ Quantified Satisfiability (QSAT)

- Quantified Boolean Formula (QBF)
- QBF evaluation and certification
- Application
 - Relation determinization, program synthesis

Satisfiability



Normal Forms

- A **literal** is a variable or its negation
- A **clause (cube)** is a disjunction (conjunction) of literals
- A **conjunctive normal form (CNF)** is a conjunction of clauses; a **disjunctive normal form (DNF)** is a disjunction of cubes

■ E.g.,

CNF: $(a + \neg b + c)(a + \neg c)(b + d)(\neg a)$

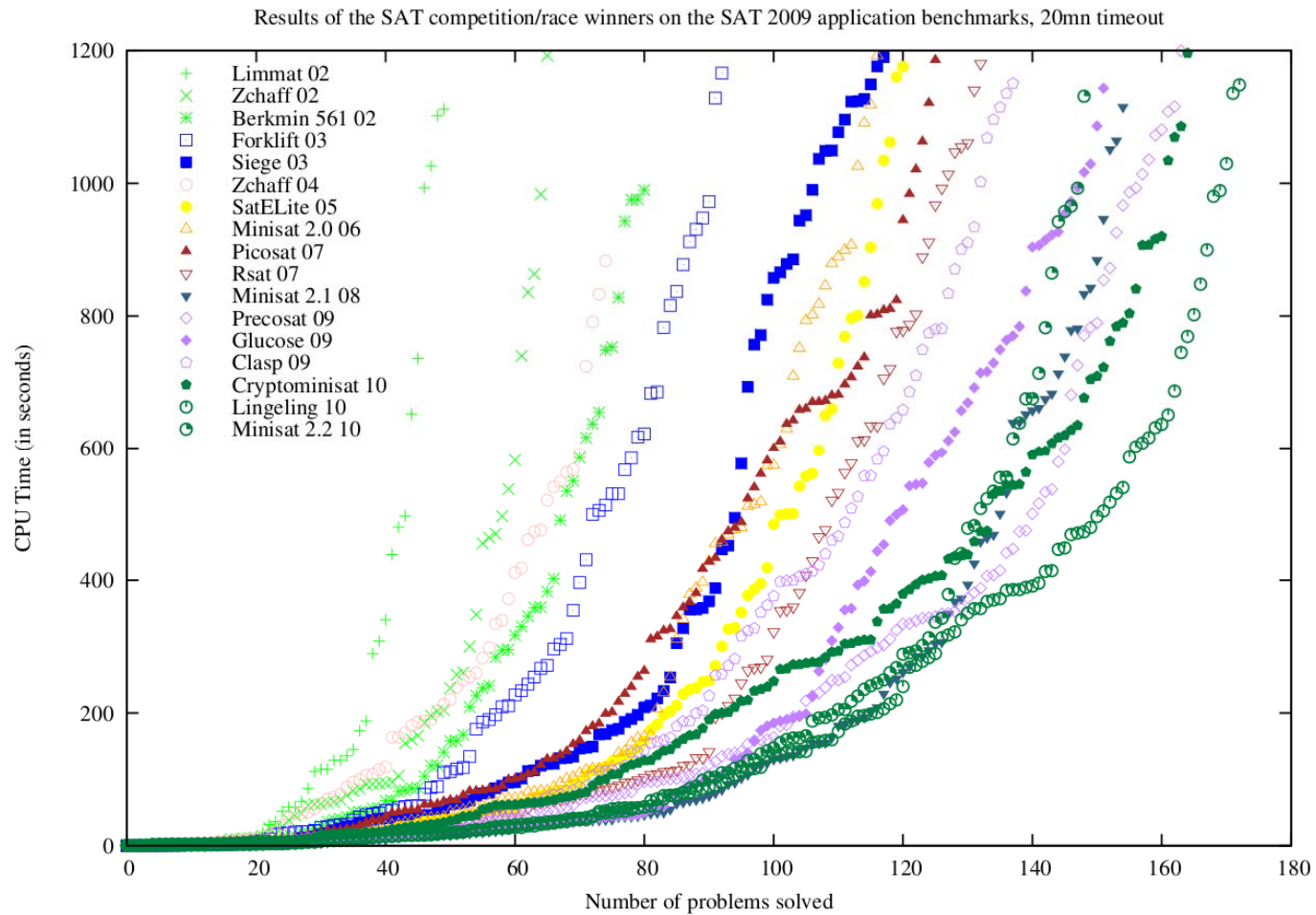
□ $(\neg a)$ is a unit clause, d is a pure literal

DNF: $a\neg bc + a\neg c + bd + \neg a$

Satisfiability

- The **satisfiability** (SAT) problem asks whether a given CNF formula can be true under some assignment to the variables
- In theory, SAT is intractable
 - The first shown NP-complete problem [Cook, 1971]
- In practice, modern SAT solvers work ‘mysteriously’ well on application CNFs with ~100,000 variables and ~1,000,000 clauses
 - It enables various applications, and inspires QBF and SMT (Satisfiability Modulo Theories) solver development

SAT Competition



<http://www.satcompetition.org/PoS11/>

SAT Solving

- ❑ Ingredients of modern SAT solvers:
 - DPLL-style search
 - ❑ [Davis, Putnam, Logemann, Loveland, 1962]
 - Conflict-driven clause learning (CDCL)
 - ❑ [Marques-Silva, Sakallah, 1996 ([GRASP](#))]
 - Boolean constraint propagation (BCP) with two-literal watch
 - ❑ [Moskewicz, Modigan, Zhao, Zhang, Malik, 2001 ([Chaff](#))]
 - Decision heuristics using variable activity
 - ❑ [Moskewicz, Modigan, Zhao, Zhang, Malik, 2001 ([Chaff](#))]
 - Restart
 - Preprocessing
 - Support for incremental solving
 - ❑ [Een, Sorensson, 2003 ([MiniSat](#))]

Pre-Modern SAT Procedure

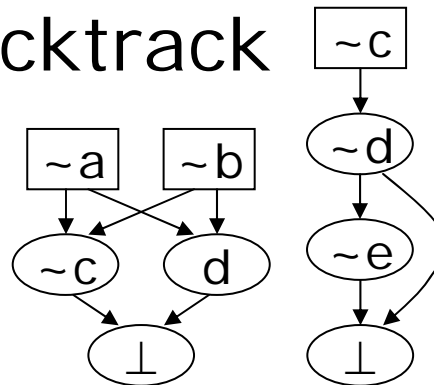
Algorithm DPLL(Φ)

```
{  
  while there is a unit clause  $\{l\}$  in  $\Phi$   
     $\Phi = \text{BCP}(\Phi, l);$   
  while there is a pure literal  $l$  in  $\Phi$   
     $\Phi = \text{assign}(\Phi, l);$   
  if all clauses of  $\Phi$  satisfied    return true;  
  if  $\Phi$  has a conflicting clause    return false;  
   $l := \text{choose\_literal}(\Phi);$   
  return  $\text{DPLL}(\text{assign}(\Phi, \neg l)) \vee \text{DPLL}(\text{assign}(\Phi, l));$   
}
```

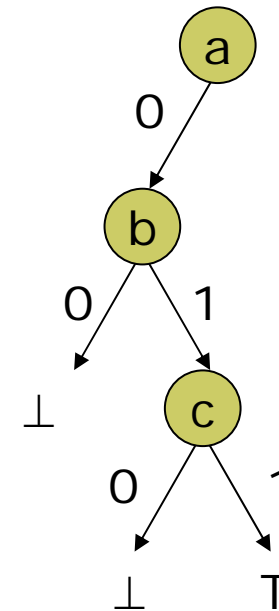

DPLL Procedure

□ Chronological backtrack

□ E.g.



	$\sim a$	$\sim b$	b	$\sim c$	c	d
$\{\neg a, e\}$	■	■	■	■	■	■
$\{a, b, \neg c\}$	□	□	■	■	■	■
$\{c, \neg d\}$	□	■	□	□	■	■
$\{a, b, d\}$	□	□	■	■	■	■
$\{d, e\}$	□	□	□	■	□	■
$\{c, d, \neg e\}$	□	□	□	□	■	■



Modern SAT Procedure

Algorithm CDCL(Φ)

```
{
  while(1)
    while there is a unit clause  $\{l\}$  in  $\Phi$ 
       $\Phi = \text{BCP}(\Phi, l);$ 
    while there is a pure literal  $l$  in  $\Phi$ 
       $\Phi = \text{assign}(\Phi, l);$ 
    if  $\Phi$  contains no conflicting clause
      if all clauses of  $\Phi$  are satisfied      return true;
       $l := \text{choose\_literal}(\Phi);$ 
       $\text{assign}(\Phi, l);$ 
    else
      if conflict at top decision level      return false;
       $\text{analyze\_conflict}();$ 
       $\text{undo assignments};$ 
       $\Phi := \text{add\_conflict\_clause}(\Phi);$ 
}
```

- There can be many learnt clauses from a conflict
- Clause learning admits non-chronological backtrack
- E.g.,
 $\{\neg x_{10587}, \neg x_{10588}, \neg x_{10592}\}$
...
 $\{\neg x_{10374}, \neg x_{10582}, \neg x_{10578}, \neg x_{10373}, \neg x_{10629}\}$
...
 $\{x_{10646}, x_{9444}, \neg x_{10373}, \neg x_{10635}, \neg x_{10637}\}$



Clause Learning as Resolution

- **Resolution** of two clauses $C_1 \vee x$ and $C_2 \vee \neg x$:

$$\frac{C_1 \vee x \quad C_2 \vee \neg x}{C_1 \vee C_2}$$

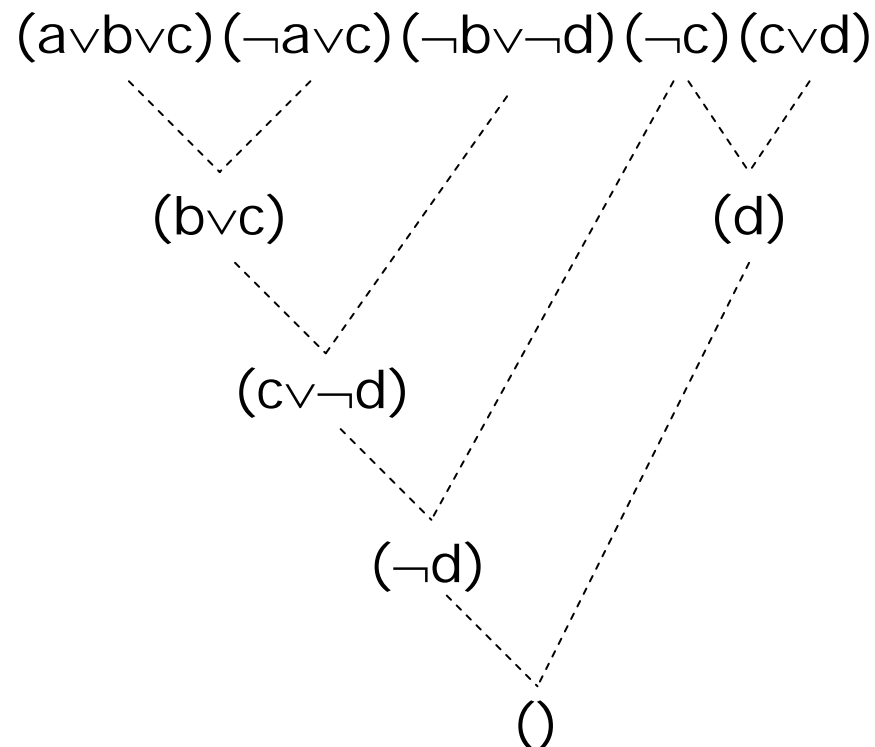
where x is the **pivot variable** and $C_1 \vee C_2$ is the **resolvent**,
i.e., $C_1 \vee C_2 = \exists x. (C_1 \vee x)(C_2 \vee \neg x)$

- A learnt clause can be obtained from a sequence of resolution steps
 - Exercise:
Find a resolution sequence leading to the learnt clause
 $\{\neg x_{10374}, \neg x_{10582}, \neg x_{10578}, \neg x_{10373}, \neg x_{10629}\}$
in the previous slides

Resolution

- Resolution is complete for SAT solving
 - A CNF formula is unsatisfiable if and only if there exists a resolution sequence leading to the empty clause

- Example



SAT Certification

□ True CNF

- Satisfying assignment (model)
 - Verifiable in linear time

□ False CNF

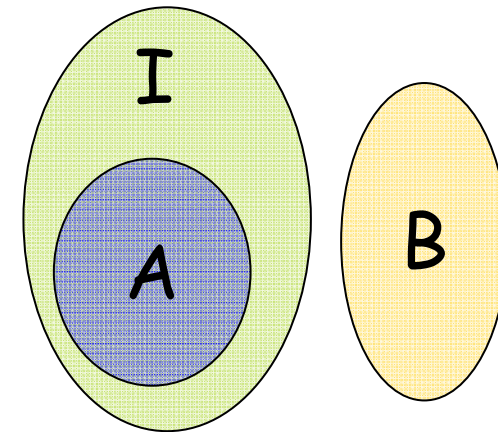
- Resolution refutation
 - Potentially of exponential size

Craig Interpolation

□ [Craig Interpolation Thm, 1957]

If $A \wedge B$ is UNSAT for formulae A and B , there exists an interpolant I of A such that

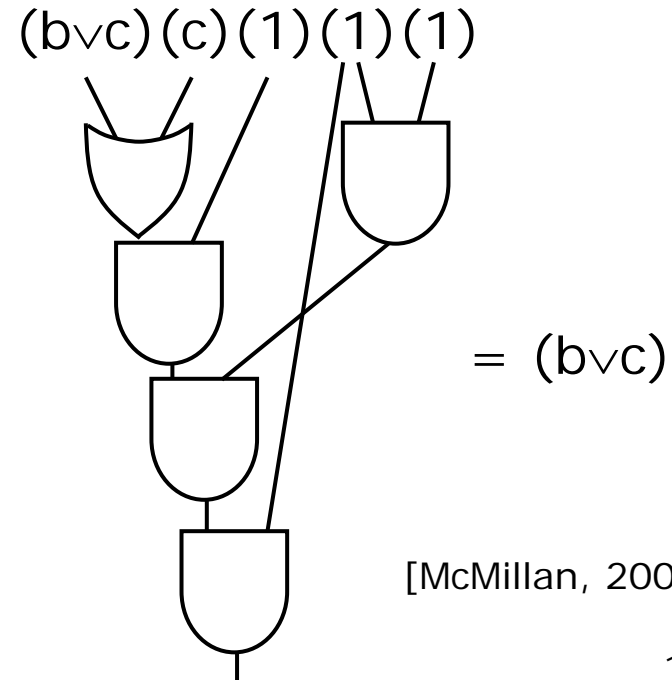
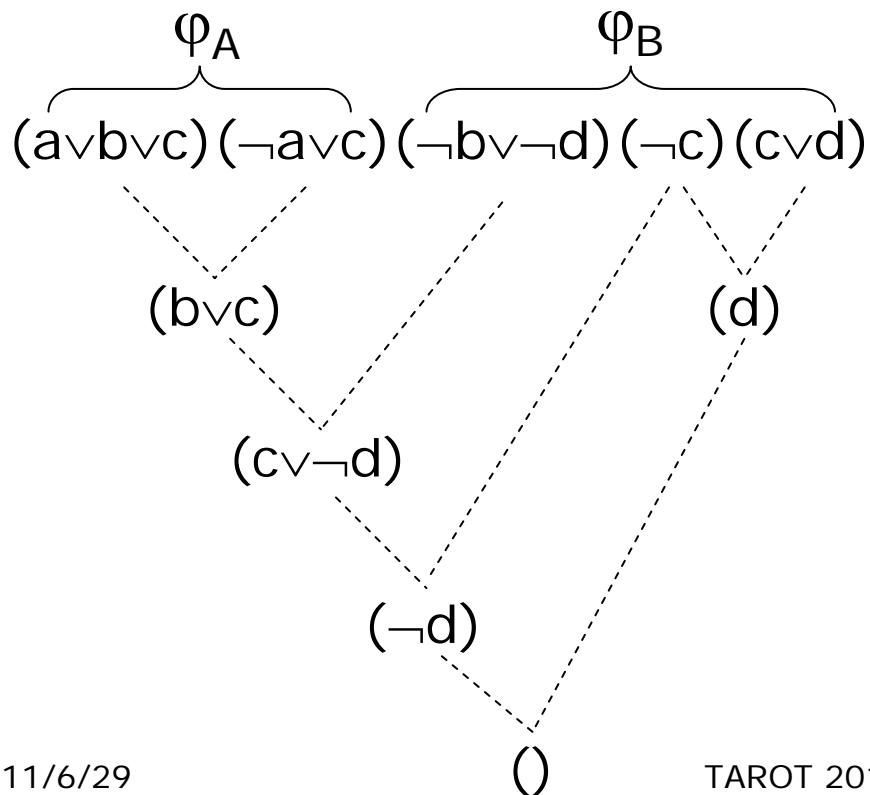
1. $A \Rightarrow I$
2. $I \wedge B$ is UNSAT
3. I refers only to the common variables of A and B



I is an abstraction of A

Interpolant and Resolution Proof

- SAT solver may produce the resolution proof of an UNSAT CNF φ
- For $\varphi = \varphi_A \wedge \varphi_B$ specified, the corresponding interpolant can be obtained in time linear in the resolution proof



[McMillan, 2003]

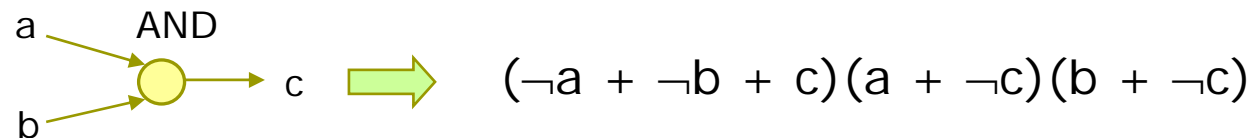
Circuit to CNF Conversion

- Circuit to CNF conversion can be done in time linear w.r.t. circuit size [Tseitin, 1968]
 - Trick: introduce intermediate variables
 - The resultant formula can blow up if no intermediate variables are allowed to exist

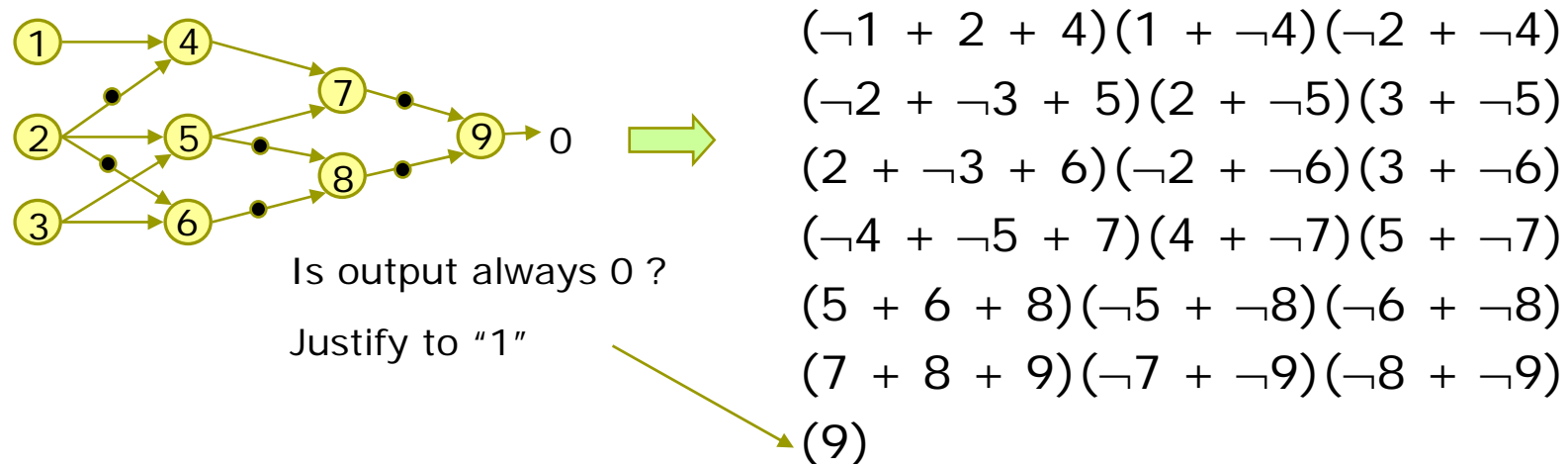
Circuit to CNF Conversion

□ Example

■ Single gate:



■ Circuit of connected gates:



SAT Application

Functional Dependency

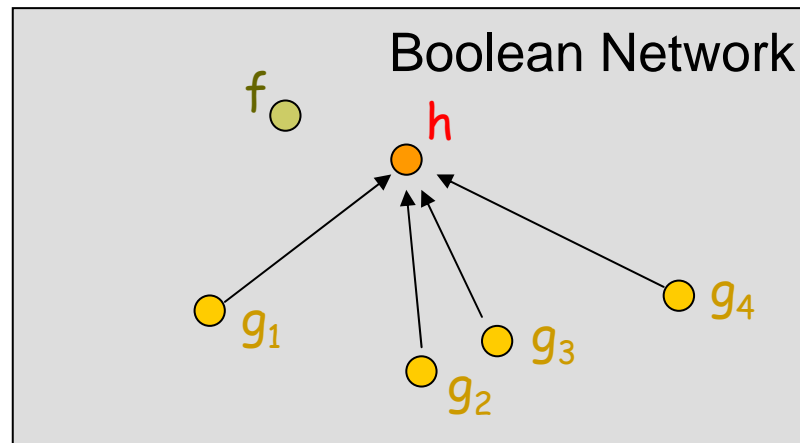
- **$f(x)$ functionally depends** on $g_1(x), g_2(x), \dots, g_m(x)$ if $f(x) = h(g_1(x), g_2(x), \dots, g_m(x))$, denoted $h(G(x))$
 - Under what condition can function f be expressed as some function h over a set of given functions $G = \{g_1, \dots, g_m\}$?
 - h exists $\Leftrightarrow \nexists a, b$ such that $f(a) \neq f(b)$ and $G(a) = G(b)$
i.e., G is more distinguishing than f

SAT Application

Functional Dependency

□ Applications of functional dependency

- Resynthesis/rewiring
- Redundant register removal
- BDD minimization
- Verification reduction
- ...



- target function
- base functions

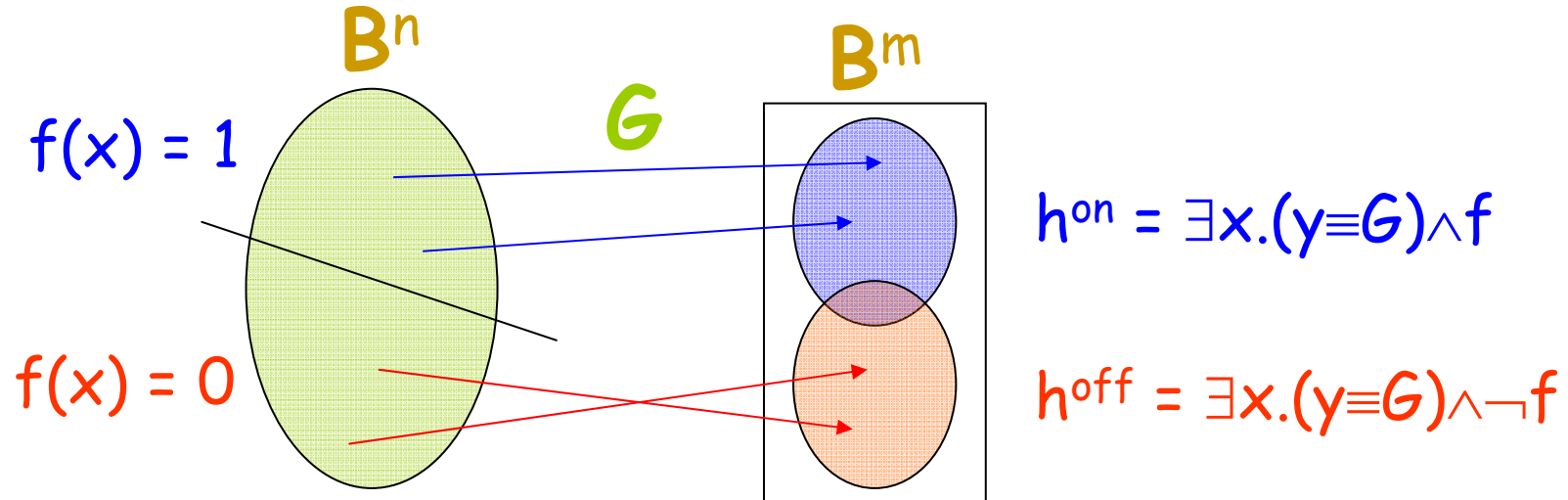
SAT Application

Functional Dependency

□ Computing h

$$h^{on} = \{y \in \mathbf{B}^m : y = G(x) \text{ and } f(x) = 1, x \in \mathbf{B}^n\}$$

$$h^{off} = \{y \in \mathbf{B}^m : y = G(x) \text{ and } f(x) = 0, x \in \mathbf{B}^n\}$$



SAT Application

Functional Dependency

□ h exists \Leftrightarrow

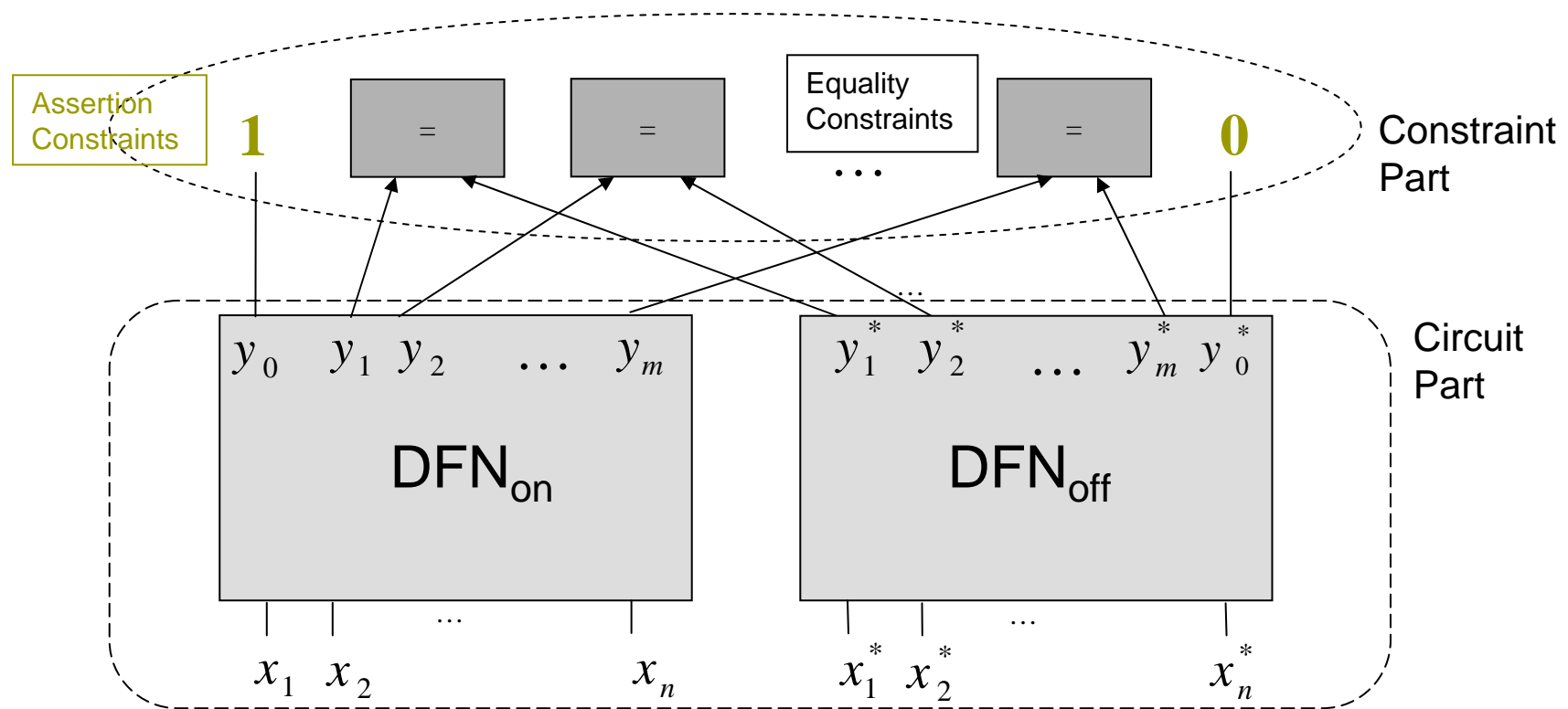
$\nexists a, b$ such that $f(a) \neq f(b)$ and $G(a) = G(b)$,
i.e., $(f(x) \neq f(x^*)) \wedge (G(x) = G(x^*))$ is **UNSAT**

□ How to derive h ? How to select G ?

SAT Application

Functional Dependency

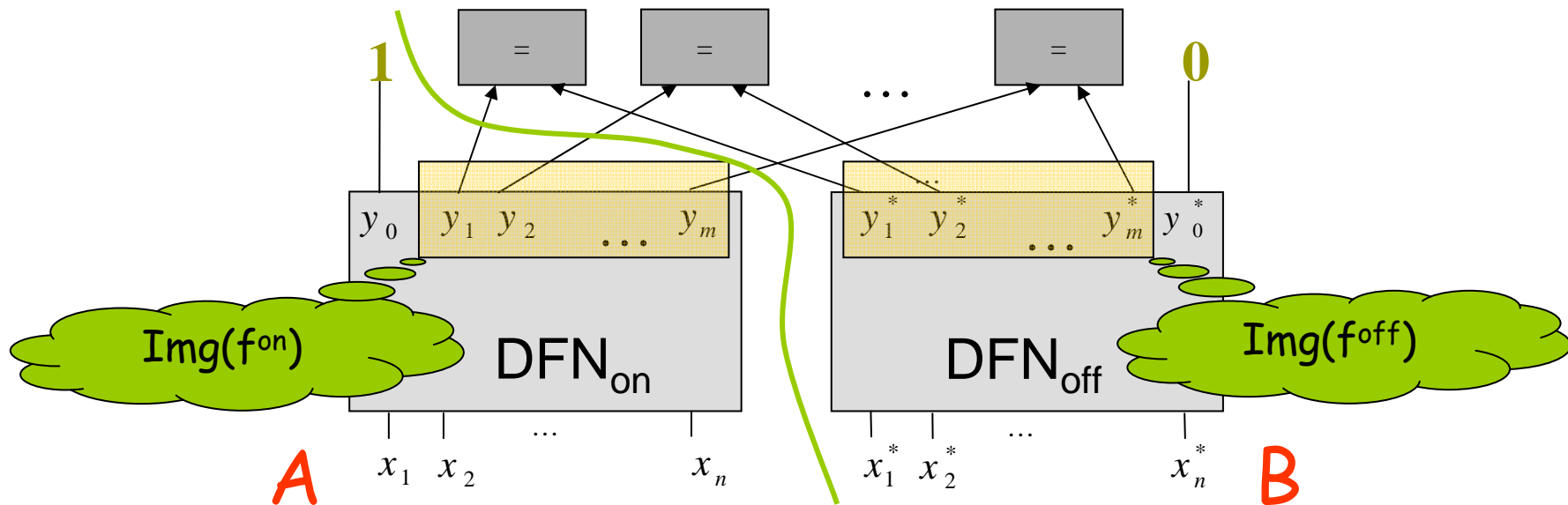
□ $(f(x) \neq f(x^*)) \wedge (G(x) \equiv G(x^*))$ is **UNSAT**



SAT Application

Functional Dependency

- Clause set **A**: $\mathcal{C}_{\text{DFN}_{\text{on}}}, y_0$
- Clause set **B**: $\mathcal{C}_{\text{DFN}_{\text{off}}}, \neg y_0^*, (y_i \equiv y_i^*)$ for $i=1, \dots, m$
- **I** is an overapproximation of $\text{Img}(f^{\text{on}})$ and is disjoint from $\text{Img}(f^{\text{off}})$
- **I** only refers to y_1, \dots, y_m
- Therefore, **I** corresponds to a feasible implementation of **h**



[Lee, J, Huang, Mishchenko, 2007]

Quantified Satisfiability



Quantified Boolean Formula

- A quantified Boolean formula (QBF) is often written in **prenex form** (with quantifiers placed on the left) as

$$\underbrace{Q_1 x_1, \dots, Q_n x_n}_{\text{prefix}} \cdot \underbrace{\varphi}_{\text{matrix}}$$

for $Q_i \in \{\forall, \exists\}$ and φ a quantifier-free formula

- If φ is further in CNF, the corresponding QBF is in the so-called **prenex CNF** (PCNF), the most popular QBF representation
- Any QBF can be converted to PCNF

Quantified Boolean Formula

- Quantification order matters in a QBF
- A variable x_i in $(Q_1 x_1, \dots, Q_i x_i, \dots, Q_n x_n. \varphi)$ is of **level** k if there are k quantifier alternations (i.e., changing from \forall to \exists or from \exists to \forall) from Q_1 to Q_i .

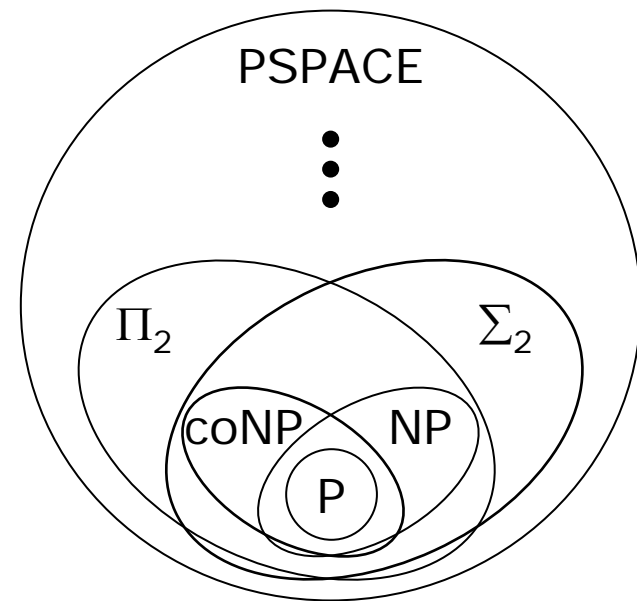
- Example

$\forall a \exists b \forall c \forall d \exists e. \varphi$

level(a)=0, level(b)=1, level(c)=2, level(d)=2,
level(e)=3

Quantified Boolean Formula

- Many decision problems can be compactly encoded in QBFs
- In theory, QBF solving (QSAT) is PSPACE complete
 - The more the quantifier alternations, the higher the complexity in the Polynomial Hierarchy
- In practice, solvable QBFs are typically of size $\sim 1,000$ variables



QBF Solver

□ QBF solver choices

- Data structures for formula representation
 - **Prenex** vs. non-prenex
 - **Normal form** vs. non-normal form
 - CNF, NNF, BDD, AIG, etc.
- Solving mechanisms
 - **Search**, Q-resolution, Skolemization, quantifier elimination, etc.
- Preprocessing techniques

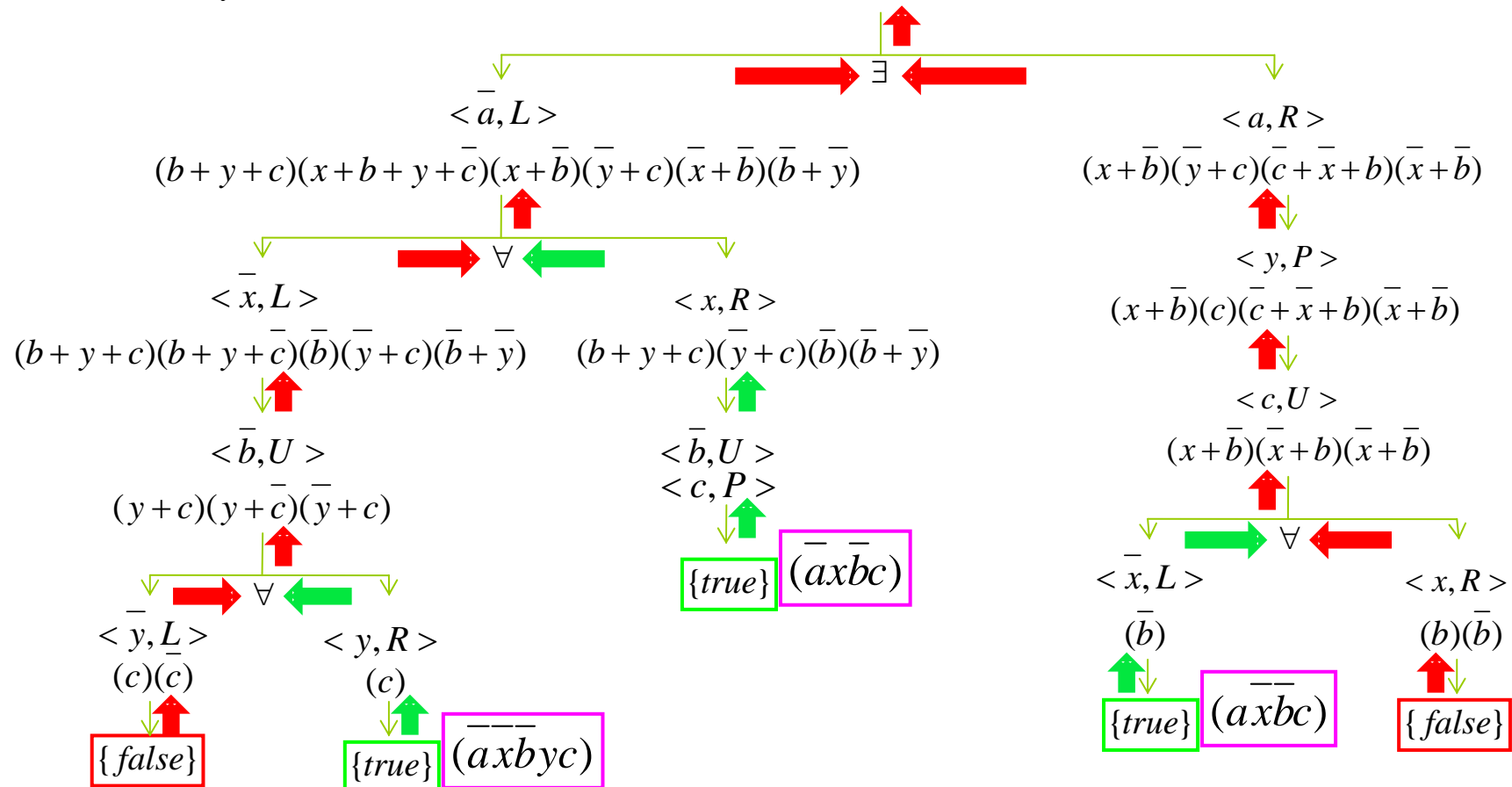
□ Standard approach

- Search-based PCNF formula solving (similar to SAT)
 - Both **clause learning** (from a conflicting assignment) and **cube learning** (from a satisfying assignment) are performed

QBF Solving

Example

$$\exists a \forall x \exists b \forall y \exists c \ (a+b+y+c)(a+x+b+y+\bar{c})(x+\bar{b})(\bar{y}+c)(\bar{c}+\bar{a}+\bar{x}+b)(\bar{x}+\bar{b})(a+\bar{b}+\bar{y})$$



Q-Resolution

- **Q-resolution** on PCNF is similar to resolution on CNF, except that the pivots are restricted to existentially quantified variables and the additional rule of **\forall -reduction**

$$\frac{C_1 \vee x \quad C_2 \vee \neg x}{\forall\text{-RED}(C_1 \vee C_2)}$$

where operator \forall -RED removes from $C_1 \vee C_2$ the universally (\forall) quantified variables whose quantification levels are greater than any of the existentially (\exists) quantified variables in $C_1 \vee C_2$

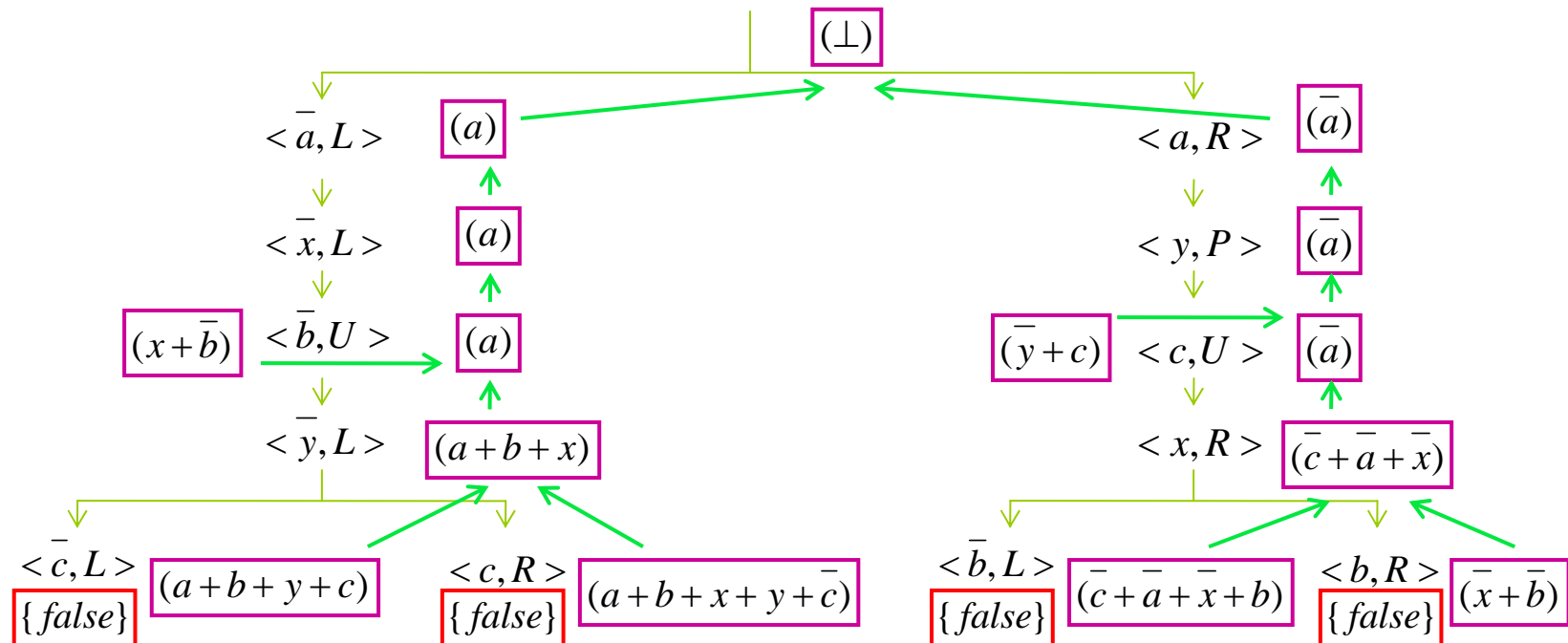
- E.g.,
prefix: $\forall a \exists b \forall c \forall d \exists e$
 $\forall\text{-RED}(a+b+c+d) = (a+b)$

- Q-resolution is complete for QBF solving
 - A PCNF formula is unsatisfiable if and only if there exists a Q-resolution sequence leading to the empty clause

Q-Resolution

Example (cont'd)

$$\exists a \forall x \exists b \forall y \exists c \quad (a + b + y + c)(a + x + b + y + \bar{c})(x + \bar{b})(\bar{y} + c)(\bar{c} + \bar{a} + \bar{x} + b)(\bar{x} + \bar{b})(a + \bar{b} + \bar{y})$$



Skolemization

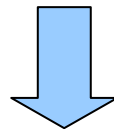
□ Skolemization and Skolem normal form

- Existentially quantified variables are replaced with function symbols
- QBF prefix contains only two quantification levels
 - \exists function symbols, \forall variables

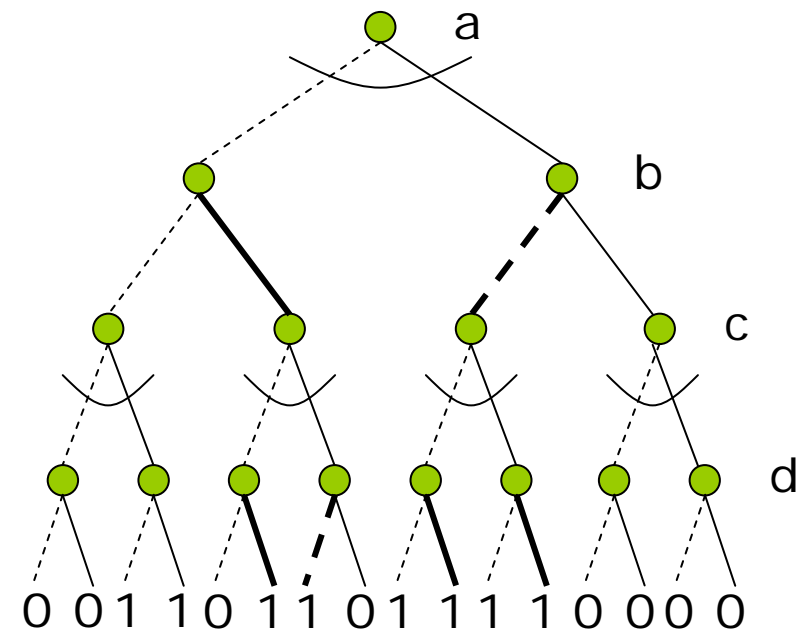
□ Example

$$\forall a \exists b \forall c \exists d. (\neg a + \neg b)(\neg b + \neg c + \neg d)(\neg b + c + d)(a + b + c)$$

Skolem functions



$$\exists F_b(a) \exists F_d(a, c) \forall a \forall c. (\neg a + \neg F_b)(\neg F_b + \neg c + \neg F_d)(\neg F_b + c + F_d)(a + F_b + c)$$



QBF Certification

□ QBF certification

- Ensure correctness and, more importantly, provide useful information
- Certificates
 - True QBF: term-resolution proof / Skolem-function (SF) model
 - SF model is more useful in practical applications
 - False QBF: clause-resolution proof / Herbrand-function (HF) countermodel
 - HF countermodel is more useful in practical applications

□ Solvers and certificates

- To date, only Skolemization-based solvers (e.g., sKizzo, squolem, Ebddres) can provide SFs
- Search-based solvers (e.g., QuBE) are the most popular and can be instrumented to provide resolution proofs

QBF Certification

□ Solvers and certificates

Solver	Algorithm	Certificate	
		True QBF	False QBF
QuBE-cert	search	Cube resolution	Clause resolution
yQuaffle	search	Cube resolution	Clause resolution
Ebddres	Skolemization	Skolem function	Clause resolution
sKizzo	Skolemization	Skolem function	-
squolem	Skolemization	Skolem function	Clause resolution

QBF Certification

□ Incomplete picture of QBF certification

	Syntactic Certificate	Semantic Certificate
True QBF	Cube-resolution proof	Skolem-function model
False QBF	Clause-resolution proof	?

□ Recent progress

■ Herbrand-function countermodel

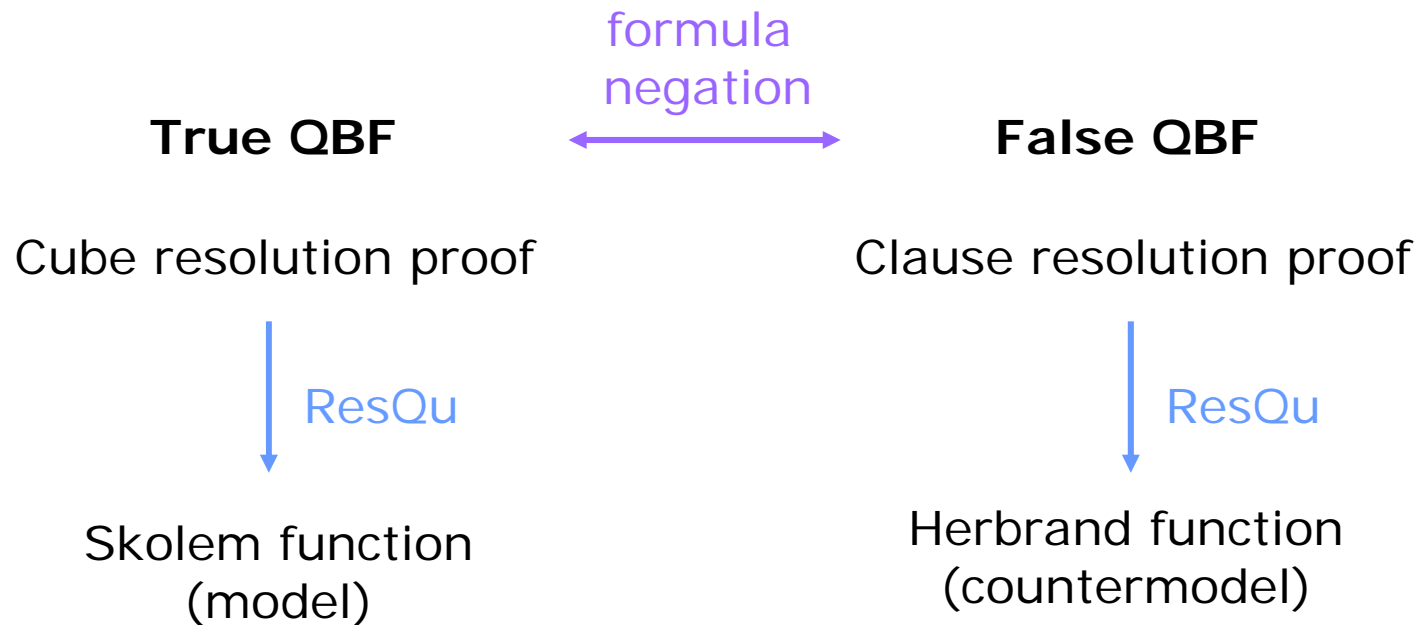
□ [Balabanov, J, 2011 ([ResQu](#))]

■ Syntactic to semantic certificate conversion

□ Linear time [Balabanov, J, 2011 ([ResQu](#))]

QBF Certification

□ Unified QBF certification



ResQu

- A Skolem-function model (Herbrand-function countermodel) for a true (false) QBF can be derived from its cube (clause) resolution proof
- A **Right-First-And-Or (RFAO) formula** is recursively defined as follows.
 $\varphi := \text{clause} \mid \text{cube} \mid \text{clause} \wedge \varphi \mid \text{cube} \vee \varphi$
 - E.g.,
 $(a'+b) \wedge ac \vee (b'+c') \wedge bc$
 $= ((a'+b) \wedge (ac \vee ((b'+c') \wedge bc)))$

ResQu

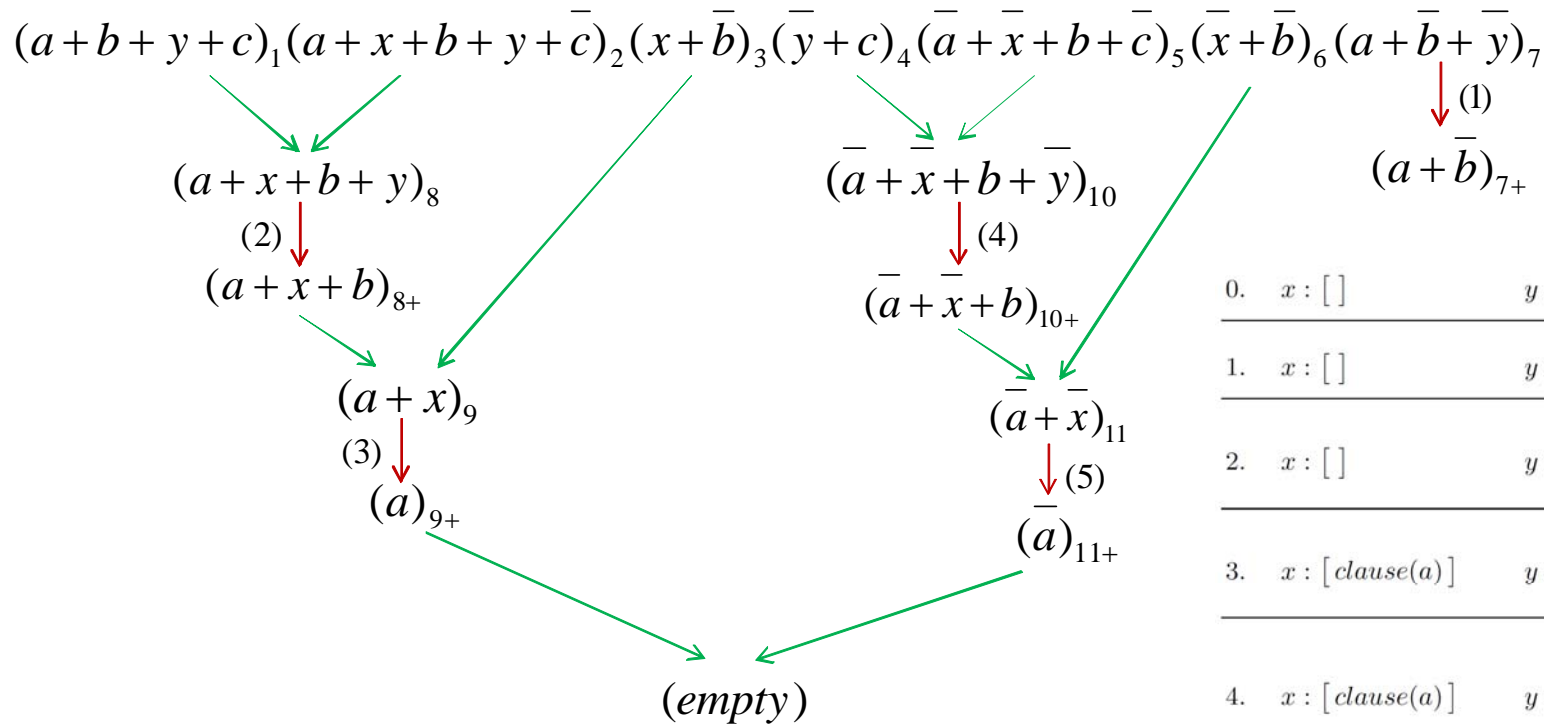
Countermodel_construct

input: a false QBF Φ and its clause-resolution DAG $G_\Pi(V_\Pi, E_\Pi)$
output: a countermodel in RFAO formulas
begin
01 **foreach** universal variable x of Φ
02 RFAO_node_array[x] := \emptyset ;
03 **foreach** vertex v of G_Π in topological order
04 **if** $v.clause$ resulted from \forall -reduction on $u.clause$, i.e., $(u, v) \in E_\Pi$
05 $v.cube := \neg(v.clause)$;
06 **foreach** universal variable x reduced from $u.clause$ to get $v.clause$
07 **if** x appears as positive literal in $u.clause$
08 **push** $v.clause$ to RFAO_node_array[x];
09 **else if** x appears as negative literal in $u.clause$
10 **push** $v.cube$ to RFAO_node_array[x];
11 **if** $v.clause$ is the empty clause
12 **foreach** universal variable x of Φ
13 simplify RFAO_node_array[x];
14 **return** RFAO_node_array's;
end

ResQu

□ Example

■ $\exists a \forall x \exists b \forall y \exists c$



0.	$x : []$	$y : []$
1.	$x : []$	$y : [cube(\bar{a}b)]$
2.	$x : []$	$y : [cube(\bar{a}b), clause(a+x+b)]$
3.	$x : [clause(a)]$	$y : [cube(\bar{a}b), clause(a+x+b)]$
4.	$x : [clause(a)]$	$y : [cube(\bar{a}b), clause(a+x+b), cube(ax\bar{b})]$
5.	$x : [clause(a), cube(a)]$	$y : [cube(\bar{a}b), clause(a+x+b), cube(ax\bar{b})]$

QBF Certification

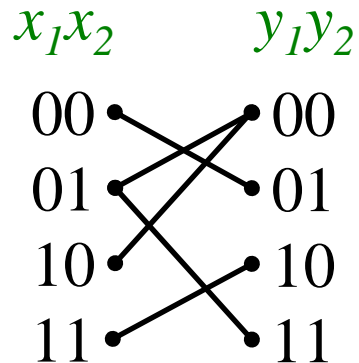
- Applications of Skolem/Herbrand functions
 - Program synthesis
 - Winning strategy synthesis in two player games
 - Plan derivation in AI
 - Logic synthesis
 - ...

QBF Application

Relation Determinization

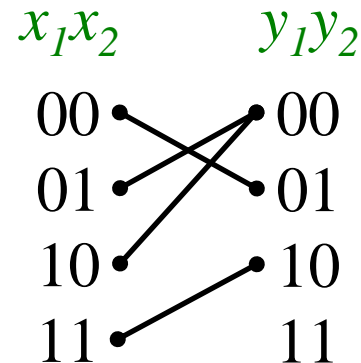
□ Relation $R(X, Y)$

- Allow one-to-many mappings
 - Can describe non-deterministic behavior
- More generic than functions



□ Function $F(X)$

- Disallow one-to-many mappings
 - Can only describe deterministic behavior
- A special case of relation



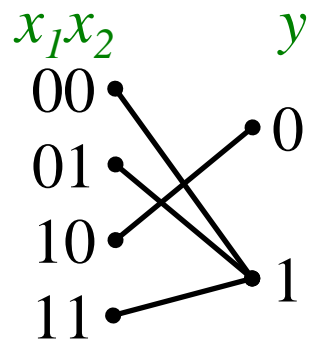
$$f_1 = x_1 x_2$$
$$f_2 = \neg x_1 \neg x_2$$

QBF Application

Relation Determinization

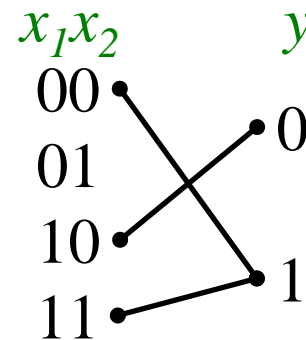
□ Total relation

- Every input element is mapped to at least one output element



□ Partial relation

- Some input element is not mapped to any output element

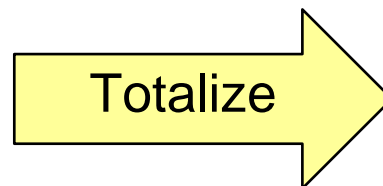
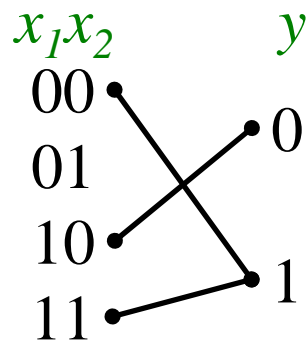


QBF Application

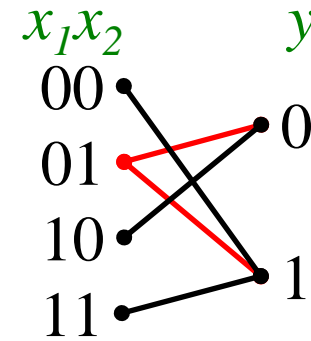
Relation Determinization

- A partial relation can be **totalized**
 - Assume that the input element not mapped to any output element is a don't care

Partial relation



Total relation

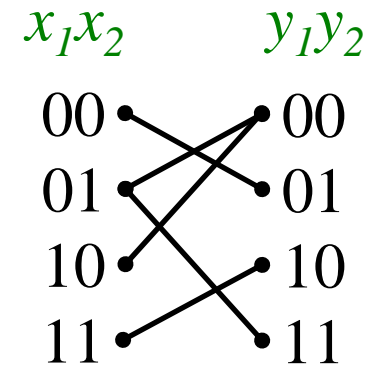
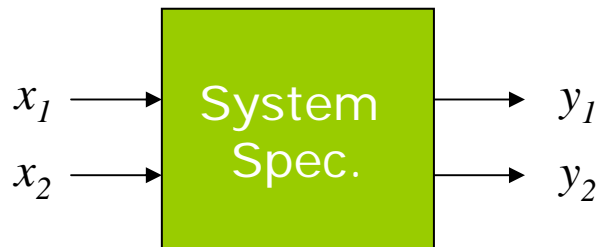


$$T(X, y) = R(X, y) \vee \forall y. \neg R(X, y)$$

QBF Application Relation Determinization

□ Applications of Boolean relation

- In high-level design, Boolean relations can be used to describe (nondeterministic) specifications
- In gate-level design, Boolean relations can be used to characterize the flexibility of sub-circuits
 - Boolean relations are more powerful than traditional don't-care representations



QBF Application

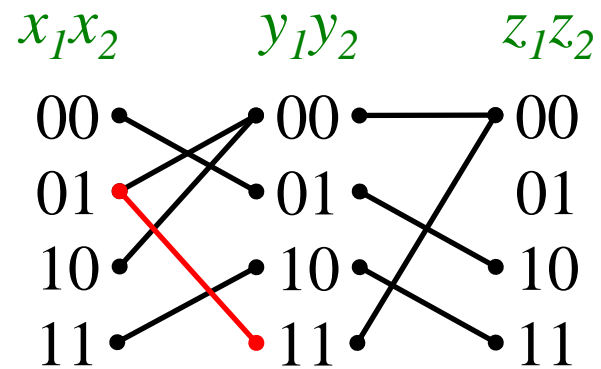
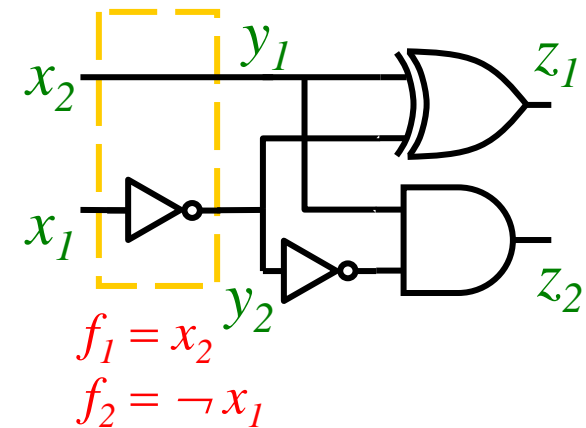
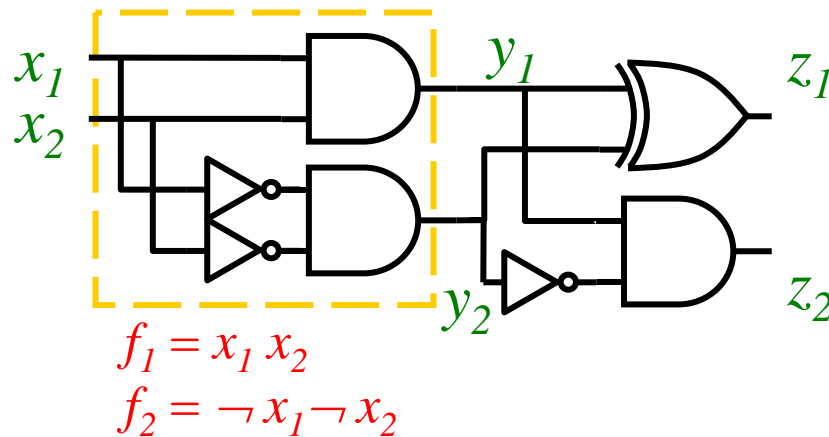
Relation Determinization

□ Relation determinization

- For hardware implementation of a system, we need functions rather than relations
 - Hardware systems are intrinsically deterministic
 - One input stimulus results in one output response
- To simplify implementation, we can explore the flexibilities described by a relation for optimization

QBF Application Relation Determinization

Example



QBF Application

Relation Determinization

□ Given a *nondeterministic* Boolean relation $R(X, Y)$, how to determinize and extract functions from it?

□ Solve QBF

$$\forall x_1, \dots, \forall x_m, \exists y_1, \dots, \exists y_n. R(x_1, \dots, x_m, y_1, \dots, y_n)$$

■ The Skolem functions of variables y_1, \dots, y_n correspond to the output-functions we want

QBF Application Program Synthesis

- Program synthesis by sketching
 - [Solar-Lezama et al., 2006]

□ Example

Spec:

```
int foo (int x){  
    return x+x;  
}
```

Sketch:

```
int bar (int x) implements foo{  
    return x << ??;  
}
```

Result:

```
int bar (int x) implements foo{  
    return x << 1;  
}
```

QBF Application Program Synthesis

- Sketch synthesis can be solved by searching for control values satisfying

$$\exists c \forall x. \text{Spec}(x) = \text{Sk}(x, c)$$

- We are interested to derive the Skolem function (in this case, constant) of c

Conclusions

- ❑ Modern SAT/QSAT solvers are powerful tools for solving large-scale synthesis, verification, and other computer science problems
- ❑ Certificates of SAT/QSAT solving may be utilized to extract essential information for applications in synthesis and verification
- ❑ Understanding how solvers work helps practitioners formulate and solve real-world problems

Suggested Further Exploration

- SMT solvers and their applications in program analysis and verification

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Thank You!

□ Questions?