

Verification – Lecture 18

Symbolic Model Checking

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REVIEW

Summary of CTL model checking (1)

- CTL is a logic for formalizing properties over computation **trees**
- The expressiveness of LTL and CTL is incomparable
- Fairness constraints cannot be expressed in CTL
 - but are incorporated by adapting the CTL semantics such that quantification is over fair paths
- CTL model checking is by a recursive descent over parse tree of Φ
 - $Sat(\exists(\Phi \cup \Psi))$ is determined using a least fixed point computation
 - $Sat(\exists\Box\Phi)$ is determined by a greatest fixed point computation

Summary of CTL model checking (2)

- Time complexity of CTL model-checking $S \models \Phi$ is:
 - is linear in $|S|$ and $|\Phi|$ and linear in k for k fairness constraints
- Checking $S \models_{fair} \Phi$ is $S \models \Phi$ plus computing $Sat_{fair}(\exists \square a)$
- CTL* is more expressive than both CTL and LTL
- The CTL* model-checking problem can be solved by an appropriate combination of the CTL and the LTL model-checking algorithm
- The CTL*-model checking problem is PSPACE-complete

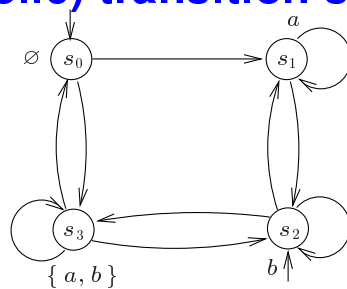
Review: Transition Systems

- finite set of **variables**: V
- **initial condition** θ : assertion over V
- finite set of **transitions** \mathcal{T}
 each $\tau \in \mathcal{T}$ represented by **transition relation** ρ_τ over $V \cup V'$
 - V : values in present state
 - V' : values in next state
- **Atomic propositions** AP : assertions over V

Boolean Transition Systems

- finite set of **boolean variables**: V
- **initial condition** θ : **boolean function** over V
- **transitions** represented by **transition relation**:
boolean function ρ over $V \cup V'$
 - V : values in present state
 - V' : values in next state
- **Atomic propositions** $AP = V$.

(Explicit) state graphs vs. (symbolic) transition systems



state	bit-vector	boolean function
s_0	$\langle 0, 0 \rangle$	$\neg x_1 \wedge \neg x_2$
s_1	$\langle 0, 1 \rangle$	$\neg x_1 \wedge x_2$
s_2	$\langle 1, 0 \rangle$	$x_1 \wedge \neg x_2$
s_3	$\langle 1, 1 \rangle$	$x_1 \wedge x_2$

- **States:**

- **Initial states:**

$$\theta(x_1, x_2) = (\neg x_1 \wedge \neg x_2) \vee (x_1 \wedge \neg x_2)$$

explicit vs. symbolic (cont'd)

- **Edge relation:**

E	$\langle 0, 0 \rangle$	$\langle 0, 1 \rangle$	$\langle 1, 0 \rangle$	$\langle 1, 1 \rangle$
$\langle 0, 0 \rangle$	0	1	0	1
$\langle 0, 1 \rangle$	0	1	1	0
$\langle 1, 0 \rangle$	0	1	1	1
$\langle 1, 1 \rangle$	1	0	1	1

- Alternatively: $\rho(\underbrace{x_1, x_2}_q, \underbrace{x'_1, x'_2}_{q'}) = 1$ if and only if $(q, q') \in E$

$$\rho(x_1, x_2, x'_1, x'_2) = \begin{aligned} & (\neg x_1 \wedge \neg x_2 \wedge \neg x'_1 \wedge x'_2) \\ \vee & (\neg x_1 \wedge \neg x_2 \wedge x'_1 \wedge x'_2) \\ \vee & (\neg x_1 \wedge x_2 \wedge x'_1 \wedge \neg x'_2) \\ \vee & \dots \\ \vee & (x_1 \wedge x_2 \wedge x'_1 \wedge x'_2) \end{aligned}$$

Boolean functions

- **Boolean functions** $f : \mathbb{B}^n \rightarrow \mathbb{B}$ for $n \geq 0$ where $\mathbb{B} = \{0, 1\}$
 - examples: $f(x_1, x_2) = x_1 \wedge (x_2 \vee \neg x_1)$, and $f(x_1, x_2) = x_1 \leftrightarrow x_2$
- **Finite sets are boolean functions**
 - let $|Q| = N$ and $2^{n-1} < N \leq 2^n$
 - each state $q \in Q$ is a boolean vector of length n : $\llbracket \cdot \rrbracket : Q \rightarrow \mathbb{B}^n$
 - $T \subseteq Q$ is represented by f_T such that:

$$f_T(\llbracket q \rrbracket) = 1 \quad \text{iff} \quad q \in T$$

- this is the **characteristic function** of T

- **Relations are boolean functions**

- $\mathcal{R} \subseteq Q \times Q$ is represented by $f_{\mathcal{R}}$ such that:

$$f_{\mathcal{R}}(\llbracket s \rrbracket, \llbracket t \rrbracket) = 1 \quad \text{iff} \quad (s, t) \in \mathcal{R}$$

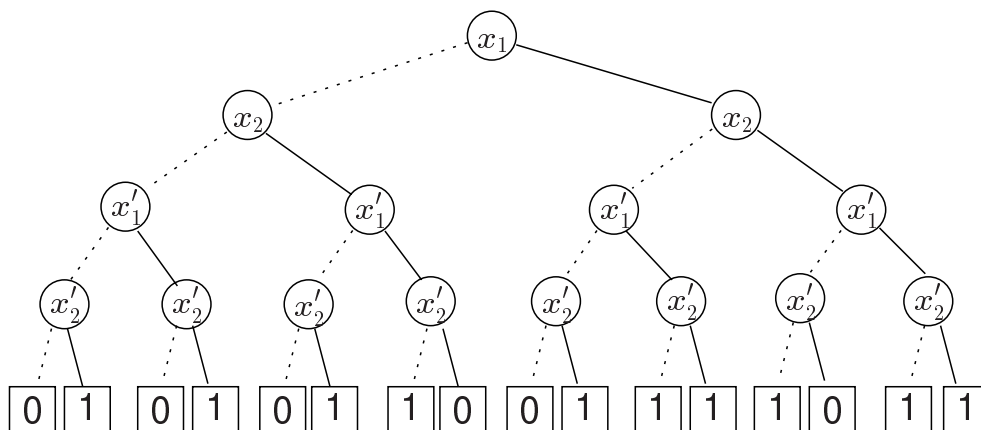
Binary decision trees

- Let X be a set of boolean variables and $<$ a total order on X
- **Binary decision tree** (BDT) is a complete binary tree over $\langle X, < \rangle$
 - each leaf v is labeled with a boolean value $val(v) \in \mathbb{B}$
 - non-leaf v is labeled by a boolean variable $Var(v) \in X$
 - such that for each non-leaf v and vertex w :

$$w \in \{ left(v), right(v) \} \Rightarrow (Var(v) < Var(w) \vee w \text{ is a leaf})$$

\Rightarrow On each path from root to leaf, variables occur in the **same order**

Transition relation as a BDT



A BDT representing ρ for our example using $x_1 < x_2 < x_1' < x_2'$

Shannon expansion

- Each boolean function $f : \mathbb{B}^n \rightarrow \mathbb{B}$ can be written as:

$$f(x_1, \dots, x_n) = (x_i \wedge f[x_i := 1]) \vee (\neg x_i \wedge f[x_i := 0])$$

- where $f[x_i := 1]$ stands for $f(x_1, \dots, x_{i-1}, 1, x_{i+1}, \dots, x_n)$
- and $f[x_i := 0]$ is a shorthand for $f(x_1, \dots, x_{i-1}, 0, x_{i+1}, \dots, x_n)$

- The boolean function $f_B(v)$ represented by vertex v in BDT B is:

- for v a leaf: $f_B(v) = \text{val}(v)$
- otherwise:

$$f_B(v) = (\text{Var}(v) \wedge f_B(\text{right}(v))) \vee (\neg \text{Var}(v) \wedge f_B(\text{left}(v)))$$

- $f_B = f_B(v)$ where v is the root of B

Considerations on BDTs

- BDTs are **not compact**
 - a BDT for boolean function $f : \mathbb{B}^b \rightarrow \mathbb{B}$ has 2^b leaves
- ⇒ they are as space inefficient as truth tables!

⇒ BDTs contain quite some **redundancy**

- all leafs with value one (zero) could be collapsed into a single leaf
- a similar scheme could be adopted for isomorphic subtrees

- The size of a BDT does not change if the variable order changes

Ordered Binary Decision Diagram

share equivalent expressions [Akers 76, Lee 59]

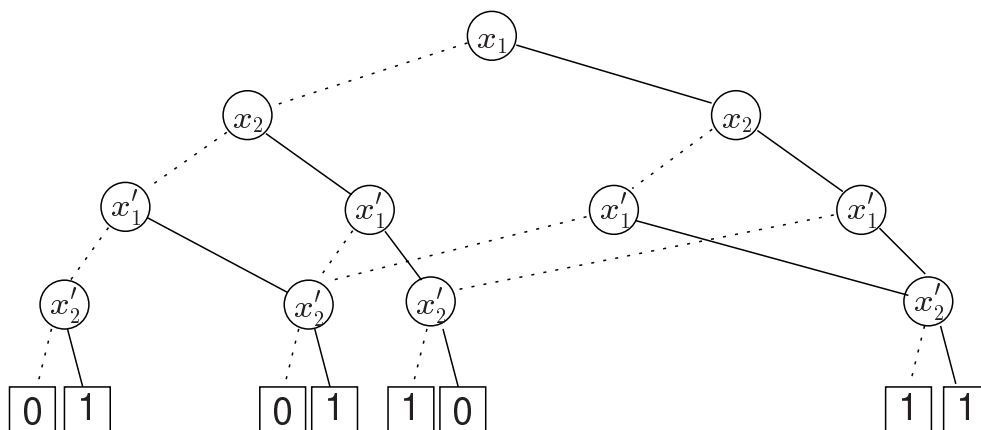
- **Binary decision diagram** (OBDD) is a **directed graph** over $\langle X, < \rangle$ with:
 - each leaf v is labeled with a boolean value $val(v) \in \{0, 1\}$
 - non-leaf v is labeled by a boolean variable $Var(v) \in X$
 - such that for each non-leaf v and vertex w :

$$w \in \{left(v), right(v)\} \Rightarrow (Var(v) < Var(w) \vee w \text{ is a leaf})$$

\Rightarrow An OBDD is acyclic

- f_B for OBDD B is obtained as for BDTs

Transition relation as an OBDD



An example OBDD representing ρ for our example using $x_1 < x_2 < x_1' < x_2'$

Isomorphism

- B and B' over $\langle X, < \rangle$ are *isomorphic* iff their roots are isomorphic
- Vertices v in B and w in B' are isomorphic, denoted $v \cong w$, iff there exists a bijection H between the vertices of B and B' such that:
 1. if v is a leaf, then $H(v) = w$ is a leaf with $val(v) = val(H(v))$
 2. if v is a non-leaf, then $H(v) = w$ is a non-leaf such that

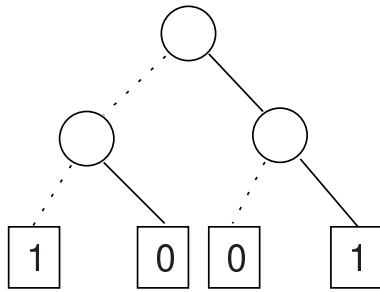
$$Var(v) = Var(w) \wedge H(left(v)) = left(H(v)) \wedge H(right(v)) = right(H(v))$$

- Testing $B \cong B'$ can be done in linear time
 - due to the labels (0 and 1) of the edges.

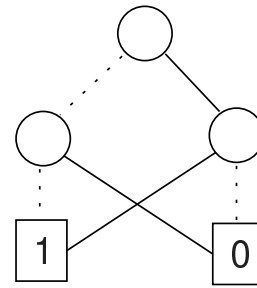
Reducing OBDDs

- Generate an OBDD (or BDT) for a boolean expression, then **reduce**
 - by means of a recursive descent over the OBDD
- **Elimination of duplicate leafs**
 - for a duplicate 0-leaf (or 1-leaf), redirect all incoming edges to just one of them
- **Elimination of “don't care” (non-leaf) vertices**
 - if $left(v) = right(v) = w$, eliminate v and redirect all its incoming edges to w
- **Elimination of isomorphic subtrees**
 - if $v \neq w$ are roots of isomorphic subtrees, remove w
 - and redirect all incoming edges to w to v

How to reduce an OBDD?

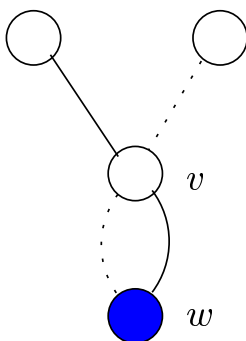


becomes

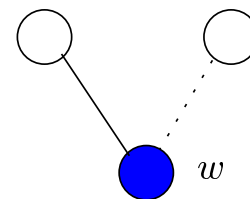


eliminating identical leafs

How to reduce an OBDD?

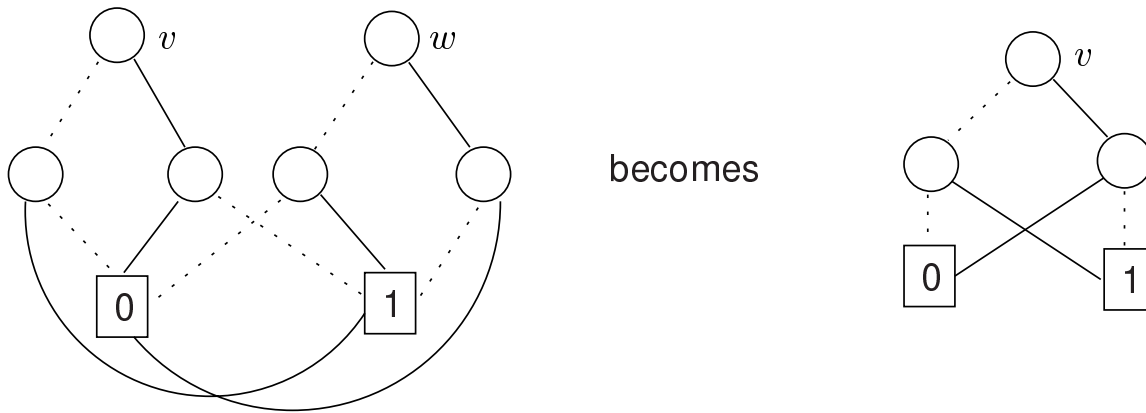


becomes



eliminating "don't care" vertices

How to reduce a BDD?



becomes

eliminating isomorphic subtrees

Reduced OBDDs

OBDD B over $\langle X, < \rangle$ is called *reduced* iff:

1. for each leaf v, w : $(val(v) = val(w)) \Rightarrow v = w$

\Rightarrow identical terminal vertices are forbidden

2. for each non-leaf v : $left(v) \neq right(v)$

\Rightarrow non-leafs may not have identical children

3. for each non-leaf v, w :

$(Var(v) = Var(w) \wedge right(v) \cong right(w) \wedge left(v) \cong left(w)) \Rightarrow v = w$

\Rightarrow vertices may not have isomorphic sub-dags

this is what is mostly called BDD; in fact it is an ROBDD!

Dynamic generation of ROBDDs

Main idea:

- Construct directly an ROBDD from a boolean expression
- Create vertices in depth-first search order
- On-the-fly reduction by applying **hashing**
 - on encountering a new vertex v , check whether:
 - an equivalent vertex w has been created (same label and children)
 - $left(v) = right(v)$, i.e., vertex v is a “don't care” vertex

ROBDDs are canonical

[Fortune, Hopcroft & Schmidt, 1978]

For ROBDDs B and B' over $\langle X, < \rangle$ we have:
 $(f_B = f_{B'})$ implies B and B' are isomorphic

\Rightarrow for a fixed variable ordering, any boolean function
can be uniquely represented by an ROBDD (up to isomorphism)

The importance of canonicity

- **Absence of redundant vertices**
 - if f_B does not depend on x_i , ROBDD B does not contain an x_i vertex
- **Test for equivalence:** $f(x_1, \dots, x_n) \equiv g(x_1, \dots, x_n)$?
 - generate ROBDDs B_f and B_g , and check isomorphism
- **Test for validity:** $f(x_1, \dots, x_n) = 1$?
 - generate ROBDD B_f and check whether it only consists of a 1-leaf
- **Test for implication:** $f(x_1, \dots, x_n) \rightarrow g(x_1, \dots, x_n)$?
 - generate ROBDD $B_f \wedge \neg B_g$ and check if it just consist of a 0-leaf
- **Test for satisfiability**
 - f is satisfiable if and only if B_f is not just the 0-leaf

Variable ordering

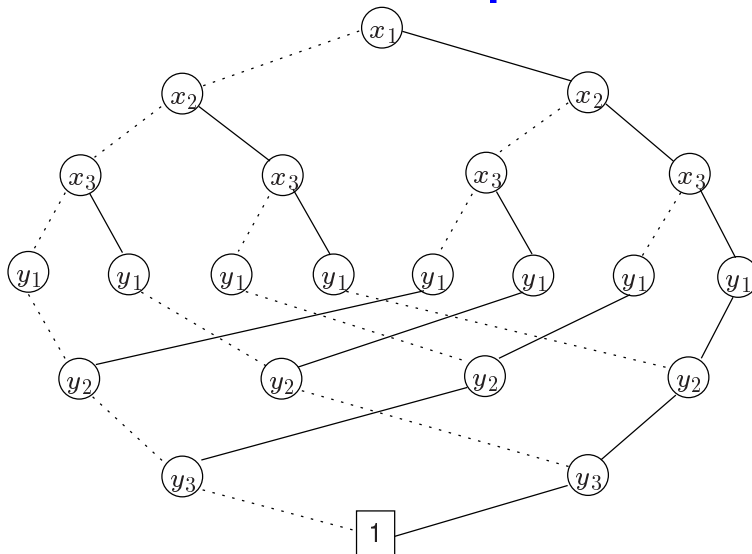
- **The size of the ROBDD depends on the variable ordering**
- **For some functions, very compact ROBDDs may be obtained**
 - e.g., the even parity function
- **Some boolean functions have linear and exponential ROBDDs**
 - e.g., the addition function, or the stable function
- **Some boolean functions only have polynomial ROBDDs**
 - this holds, e.g., for symmetric functions (see next)
 - examples $f(\dots) = x_1 \oplus \dots \oplus x_n$, or $f(\dots) = 1$ iff $\geq k$ variables x_i are true
- **Some boolean functions only have exponential ROBDDs**
 - this holds, e.g., for the multiplication function, cf. (Bryant, 1986)

The even parity function

$f_{\text{even}}(x_1, \dots, x_n) = 1$ iff the number of variables x_i with value 1 is even

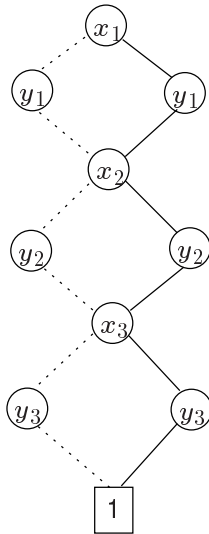
*truth table or propositional formula for f_{even} has exponential size
but an ROBDD of linear size is possible*

The function stable with exponential ROBDD



The ROBDD of $f_{\text{stab}}(\bar{x}, \bar{y}) = (x_1 \leftrightarrow y_1) \wedge \dots \wedge (x_n \leftrightarrow y_n)$
has $3 \cdot 2^n - 1$ vertices under ordering $x_1 < \dots < x_n < y_1 < \dots < y_n$

The function stable with linear ROBDD



The ROBDD of $f_{stab}(\bar{x}, \bar{y}) = (x_1 \leftrightarrow y_1) \wedge \dots \wedge (x_n \leftrightarrow y_n)$
has $3 \cdot n + 2$ vertices under ordering $x_1 < y_1 < \dots < x_n < y_n$

Symmetric functions

$$f[x_1 := b_1, \dots, x_n := b_n] = f[x_1 := b_{i_1}, \dots, x_{i_n} := b_{i_n}]$$

for each permutation (i_1, \dots, i_n) of $(1, \dots, n)$

symmetric boolean functions have ROBDDs of size in $\mathcal{O}(n^2)$

The multiplication function

- Consider two n -bit integers
 - let $b_{n-1}b_{n-2} \dots b_0$ and $c_{n-1}c_{n-2} \dots c_0$
 - where b_{n-1} is the most significant bit, and b_0 the least significant bit
- Multiplication yields a $2n$ -bit integer
 - the ROBDD $B_{f_{n-1}}$ has at least 1.09^n vertices
 - where f_{n-1} denotes the the $(n-1)$ -st output bit of the multiplication

Optimal variable ordering

- The size of ROBDDs is dependent on the variable ordering
- Is it possible to determine \prec such that the ROBDD has minimal size?
 - the optimal variable ordering problem for ROBDDs is NP-complete
 - polynomial reduction from the 3SAT problem (Bollig & Wegener, 1996)
- There are many boolean functions with large ROBDDs
 - for almost all boolean functions the minimal size is in $\Omega(\frac{2^n}{n})$
- How to deal with this problem in practice?
 - guess a variable ordering in advance
 - rearrange the variable ordering during the manipulations of ROBDDs

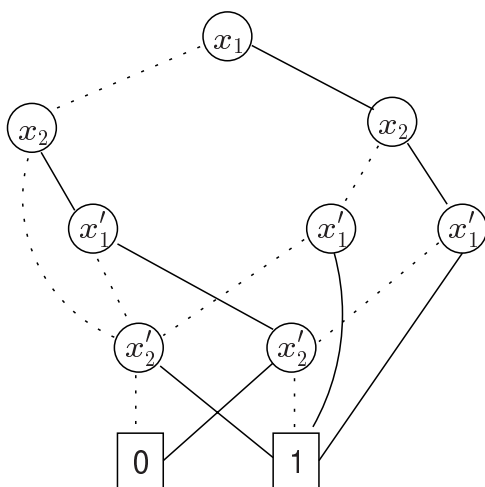
Sifting algorithm

(Rudell, 1993)

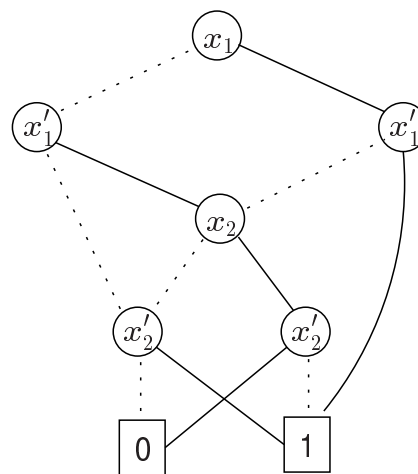
Dynamic variable ordering using variable swapping:

1. Select a variable x_i
 2. By successive swapping of x_i , determine $|B|$ at any position for x_i
 3. Shift x_i to its optimal position
 4. Go back to the first step until no improvement is made
- o Characteristics:
 - a variable may change position several times during a single sifting iteration
 - often yields a local optimum, but works well in practice

Transition relation as an ROBDD



(a) ordering $x_1 < x_2 < x'_1 < x'_2$



(b) ordering $x_1 <' x'_1 <' x_2 <' x'_2$

Interleaved variable ordering

- Which variable ordering to use for transition relations?
- The *interleaved variable ordering*:
 - for encodings x_1, \dots, x_n and y_1, \dots, y_n of state s and t respectively:

$$x_1 < y_1 < x_2 < y_2 < \dots < x_n < y_n$$

- This variable ordering yields compact ROBDDs for binary relations

Operations on ROBDDs

Algorithm	Inputs	Output ROBDD
REDUCE	B (not reduced)	B' (reduced) with $f_B = f_{B'}$
NOT	B_f	$B_{\neg f}$
APPLY	B_f, B_g , binary logical operator op	$B_f op g$
RESTRICT	B_f , variable x , boolean value b	$B_{f[x:=b]}$
RENAME	B_f , variables x and y	$B_{f[x:=y]}$
EXISTS	B_f , variable x	$B_{\exists x. f}$

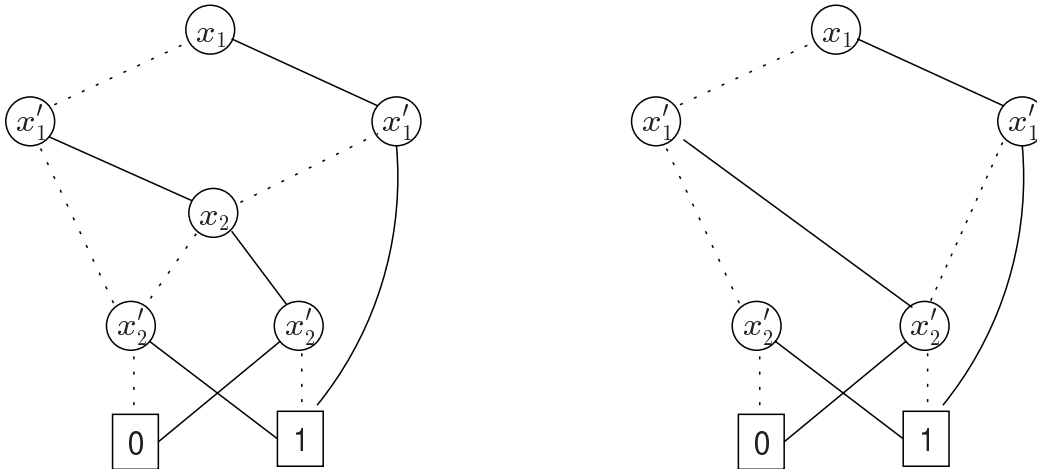
Algorithm APPLY(op, B_f, B_g)

```
B.root := APPLY( $op, B_f.root, B_g.root$ );  
  
if  $G(v_1, v_2) \neq \text{empty}$  then return  $G(v_1, v_2)$  fi;           (* lookup in hashtable *)  
if ( $v_1$  and  $v_2$  are terminals) then  $res := val(v_1) op val(v_2)$  fi;  
else if ( $v_1$  is terminal and  $v_2$  is nonterminal)  
    then  $res := MakeNode(Var(v_2), APPLY(op, v_1, left(v_2)), APPLY(op, v_1, right(v_2)))$ ;  
else if ( $v_1$  is nonterminal and  $v_2$  is terminal)  
    then  $res := MakeNode(Var(v_1), APPLY(op, left(v_1), v_2), APPLY(op, right(v_1), v_2))$ ;  
else if ( $Var(v_1) = Var(v_2)$ )  
    then  $res := MakeNode(Var(v_1), APPLY(op, left(v_1), left(v_2)), APPLY(op, right(v_1), right(v_2)))$ ;  
else if ( $Var(v_1) < Var(v_2)$ )  
    then  $res := MakeNode(Var(v_1), APPLY(op, left(v_1), v_2), APPLY(op, right(v_1), v_2))$ ;  
else  
     $res := MakeNode(Var(v_2), APPLY(op, v_1, left(v_2)), APPLY(op, v_1, right(v_2)))$ ;  
    (*  $Var(v_1) > Var(v_2)$  *)  
 $G(v_1, v_2) := res$ ;                                         (* memoize result *)  
return  $res$ 
```

Algorithm RESTRICT(B, x, b)

- For each vertex v labeled with variable x :
 - if $b = 1$ then redirect incoming edges to $right(v)$
 - if $b = 0$ then redirect incoming edges to $left(v)$
 - remove vertex v , and (if necessary) reduce (only above v)

RESTRICT



performing $\text{RESTRICT}(B, x_2, 1)$: replace x_2 by constant 1

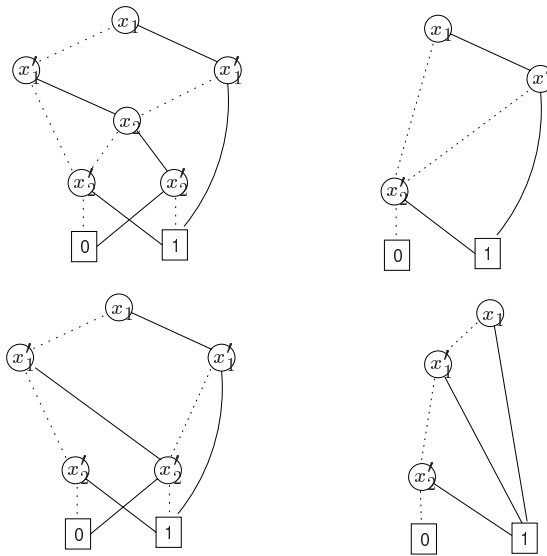
EXISTS

- Existential quantification over x_i :

$$\exists x_i. f(x_1, \dots, x_n) = f[x_i := 1] \vee f[x_i := 0]$$

- Naive realization: $\text{APPLY}(\vee, \text{RESTRICT}(B_f, x_i, 1), \text{RESTRICT}(B_f, x_i, 0))$
- Efficiency gain:
 - observe that $\text{RESTRICT}(B_f, x_i, 1)$ and $\text{RESTRICT}(B_f, x_i, 0)$ are equal up to x_i
 - ... the resulting ROBDD also has the same structure up to x_i
 - replace each node labeled with x_i by the result of applying \vee on its children
- This can easily be generalized to $\exists x_1. \dots \exists x_k. f(x_1, \dots, x_n)$

A more involved example



ROBDDs B_f (left up), $B_{f[x_2:=0]}$ (right up), $B_{f[x_2:=1]}$ (left down), and $B_{\exists x_2. f}$ (right down)

Operations on ROBDDs

Algorithm	Output	Time complexity	Space complexity
REDUCE	B' (reduced) with $f_B = f_{B'}$	$\mathcal{O}(B_f \cdot \log B_f)$	$\mathcal{O}(B_f)$
NOT	$B_{\neg f}$	$\mathcal{O}(B_f)$	$\mathcal{O}(B_f)$
APPLY	$B_{f \text{ op } g}$	$\mathcal{O}(B_f \cdot B_g)$	$\mathcal{O}(B_f \cdot B_g)$
RESTRICT	$B_{f[x:=b]}$	$\mathcal{O}(B_f)$	$\mathcal{O}(B_f)$
RENAME	$B_{f[x:=y]}$	$\mathcal{O}(B_f)$	$\mathcal{O}(B_f)$
EXISTS	$B_{\exists x. f}$	$\mathcal{O}(B_f ^2)$	$\mathcal{O}(B_f ^2)$

operations are only efficient if f and g have compact ROBDD representations