

# Tecniche di Specifica e di Verifica

Boolean Decision Diagrams I  
(BDDs)

# Outline

- **NuSMV**
- The state explosion problem.
- Techniques for overcoming this problem:
  - Compact representation of the state space.
    - **BDDs.**
  - **Abstractions (bisimulations)**
  - **Symmetries.**
  - **Partial Order Reductions.**

# NuSMV

- **N**ew **S**ymbolic **M**odel **V**erifier.
- Developed at **CMU-IRST** (**Ed Clarke, Ken McMillan, Cimatti et al.**) as extension/reimplementation of **SMV**.
- **NuSMV** has its own input language (also called **SMV!**).

# NuSMV

- You must prepare your verification problem in this language.
- An **NuSMV** program is a convenient way to describe a **Kripke structure**.
- You can insert the properties you want to verify in the program.
- Read the tutorial and on a need-to-know basis, the manual.

# Parallel Composition

- $\mathbf{TS}_1 = (\mathbf{S}_1, \mathbf{S}_1^0, \Sigma_1, \mathbf{R}_1)$     $\mathbf{R}_1 \subseteq \mathbf{S}_1 \times \Sigma_1 \times \mathbf{S}_1$
- $\mathbf{TS}_2 = (\mathbf{S}_2, \mathbf{S}_2^0, \Sigma_2, \mathbf{R}_2)$     $\mathbf{R}_2 \subseteq \mathbf{S}_2 \times \Sigma_2 \times \mathbf{S}_2$

- $\mathbf{a} \in \Sigma_1$  and  $\mathbf{a} \notin \Sigma_2$ 
  - An “*internal*” action of  $\mathbf{TS}_1$ .
- $\mathbf{a} \in \Sigma_1 \cap \Sigma_2$ 
  - A *common (synchronizing)* action of  $\mathbf{TS}_1$  and  $\mathbf{TS}_2$ .

# Parallel Composition

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- $\mathbf{TS}_2 = (\mathbf{S}_2, \mathbf{S}_2^0, \Sigma_2, \mathbf{R}_2)$     $\mathbf{R}_2 \subseteq \mathbf{S}_2 \times \Sigma_2 \times \mathbf{S}_2$
- $\mathbf{TS} = (\mathbf{TS}_1 \parallel \mathbf{TS}_2) = (\mathbf{S}, \mathbf{S}^0, \Sigma, \mathbf{R})$ .

- $\mathbf{S} = \mathbf{S}_1 \times \mathbf{S}_2$
- $\mathbf{S}^0 = \mathbf{S}_1^0 \times \mathbf{S}_2^0$
- $\Sigma = \Sigma_1 \cup \Sigma_2$

# Parallel Composition

- $\mathbf{TS}_1 = (\mathbf{S}_1, \mathbf{S}_1^0, \Sigma_1, \mathbf{R}_1)$   $\mathbf{R}_1 \subseteq \mathbf{S}_1 \times \Sigma_1 \times \mathbf{S}_1$
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- $\mathbf{TS} = (\mathbf{TS}_1 \parallel \mathbf{TS}_2) = (\mathbf{S}, \mathbf{S}^0, \Sigma, \mathbf{R})$ .

- $\mathbf{R} \subseteq \mathbf{S} \times \Sigma \times \mathbf{S}$ 
  - $\mathbf{S} = \mathbf{S}_1 \times \mathbf{S}_2$ .
- $\mathbf{R}((s1, s2), a, (t1, t2))$  ?
- if  $a \in \Sigma_1$  and  $a \notin \Sigma_2$
- then  $\mathbf{R}_1(s1, a, t1)$  and  $s2 = t2$ .

# Parallel Composition

- $\text{TS}_1 = (\mathbf{S}_1, \mathbf{S}_1^0, \Sigma_1, \mathbf{R}_1)$   $\mathbf{R}_1 \subseteq \mathbf{S}_1 \times \Sigma_1 \times \mathbf{S}_1$
- $\text{TS}_2 = (\mathbf{S}_2, \mathbf{S}_2^0, \Sigma_2, \mathbf{R}_2)$   $\mathbf{R}_2 \subseteq \mathbf{S}_2 \times \Sigma_2 \times \mathbf{S}_2$
- $\text{TS} = (\text{TS}_1 \parallel \text{TS}_2) = (\mathbf{S}, \mathbf{S}^0, \Sigma, \mathbf{R})$ .

- $\mathbf{R} \subseteq \mathbf{S} \times \Sigma \times \mathbf{S}$ 
  - $\mathbf{S} = \mathbf{S}_1 \times \mathbf{S}_2$ .
- $\mathbf{R}((s1, s2), \mathbf{a}, (t1, t2))$  ?
- if  $\mathbf{a} \in \Sigma_2$  and  $\mathbf{a} \notin \Sigma_1$
- then  $\mathbf{R}_2(s2, \mathbf{a}, t2)$  and  $s1 = t1$ .



# Parallel Composition

- $\mathbf{TS}_1 = (\mathbf{S}_1, \mathbf{S}_1^0, \Sigma_1, \mathbf{R}_1)$     $\mathbf{R}_1 \subseteq \mathbf{S}_1 \times \Sigma_1 \times \mathbf{S}_1$
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–  $\mathbf{R} \subseteq \mathbf{S} \times \Sigma \times \mathbf{S}$

▪  $\mathbf{S} = \mathbf{S}_1 \times \mathbf{S}_2$ .

–  $\mathbf{R}((s1, s2), a, (t1, t2))$  ?

– if  $a \in \Sigma_1$  and  $a \in \Sigma_2$

– then  $\mathbf{R}_1(s1, a, t1)$  and  $\mathbf{R}_2(s2, a, t2)$

# Parallel Composition

- $\mathbf{TS} = (\mathbf{TS}_1 \parallel \mathbf{TS}_2) \parallel \mathbf{TS}_3$
- $\mathbf{TS} = \mathbf{TS}_1 \parallel (\mathbf{TS}_2 \parallel \mathbf{TS}_3)$
- $\mathbf{TS} = \mathbf{TS}_1 \parallel \mathbf{TS}_2 \parallel \mathbf{TS}_3$

# Parallel Composition

- $TS = TS_1 \parallel TS_2 \dots \parallel TS_n$
- $\text{Size}(TS_i) \approx |S_i| = k_i \geq 2$
- **Description of  $TS \approx k_1 + k_2 \dots + k_n$**
- **$\text{Size}(TS) = k_1 \times k_2 \dots \times k_n$   
 $\geq 2^n !$**
- Size of **TS** is *exponential* in **n** (the *number of components*).
- *State space explosion problem.*

# How to circumvent state space explosion?

- Use succinct representations of the state space.
  - **Boolean Decision Diagrams.**
- Reduce **TS** to **TS'** such that:
  - **TS** has the required property *iff* **TS'** has the required property.
    - Symmetries
    - **Abstractions (bisimulations)**
    - Partial order reductions.

# Symbolic Model checking

- $\mathbf{K} = (\mathbf{S}, \mathbf{S}_0, \mathbf{R}, \mathbf{AP}, \mathbf{V})$
- $\psi$  a **CTL** formula
- To check whether:
  - $\mathbf{K}, \mathbf{s} \models \psi$
- We need to

- compute  $\llbracket \psi \rrbracket = \mathbf{states}(\psi) = \{\mathbf{s} \mid \mathbf{K}, \mathbf{s} \models \psi\}$ .
- then check whether  $\mathbf{s} \in \llbracket \psi \rrbracket$ .

# Symbolic Model checking

- $\mathbf{K} = (\mathbf{S}, \mathbf{S}_0, \mathbf{R}, \mathbf{AP}, \mathbf{V})$
- $\psi$  a CTL formula

- $\mathbf{S}' \subseteq \mathbf{S}$  can be represented as a *boolean function*.
- $\mathbf{R}$  can be represented as a *boolean function*.
- $\llbracket \psi \rrbracket$  can then be represented as a *boolean function*.

- *Boolean functions* represent the *characteristic functions* of the given *sets of states*.

# BDDs

- Boolean functions can be (often) *succinctly represented* as *boolean decision diagrams*.
- **BDDs** are easy to manipulate.
- *Not all boolean functions have a succinct representation.*
- *Use **BDDs** to represent and manipulate the boolean functions associated with the model checking process.*

# Boolean Functions

- **f** : Domain  $\rightarrow$  Range
- Boolean function:
  - Domain =  $\{0, 1\}^n = \{0,1\} \times \dots \times \{0,1\}$ .
  - Range =  $\{0, 1\}$
  - **f** is a function of **n** boolean variables.
- How many boolean functions of **3** variables are there?



# Boolean Functions

- **f** : Domain  $\rightarrow$  Range
- Boolean function:
  - Domain =  $\{0, 1\}^n = \{0,1\} \times \dots \times \{0,1\}$ .
  - Range =  $\{0, 1\}$
  - **f** is a function of **n** boolean variables.
- How many boolean functions of **3** variables are there?
  - Answer :  $2^{2^3} = 2^8$  !

# Truth Tables

<b>x</b>	<b>y</b>	<b>z</b>	<b>g</b>
0	0	0	0
0	0	1	1
0	1	0	1
0	1	1	0
1	0	0	1
1	0	1	0
1	1	0	0
1	1	1	1

$$g : \{0, 1\} \times \{0, 1\} \times \{0, 1\} \rightarrow \{0, 1\}$$

# Boolean Expressions

- Given a set of *Boolean variables*  $x, y, \dots$  and the constants **1** (true) and **0** (false):

$$t ::= x \mid 0 \mid 1 \mid \neg t \mid t \wedge t \mid t \vee t \mid t \Rightarrow t \mid t \Leftrightarrow t$$

- The semantics of *Boolean Expressions* is defined by means of *truth tables* as usual.
- Given an ordering of Boolean variables, *Boolean expressions* can be used to express *Boolean functions*.

# Boolean expressions

- Boolean functions can also be represented as boolean (propositional) expressions.
- $x \wedge y$  represents the function:
  - $f : \{0, 1\} \times \{0, 1\} \rightarrow \{0, 1\}$ 
    - $f(0, 0) =$
    - $f(0, 1) =$
    - $f(1, 0) =$
    - $f(1, 1) =$

# Boolean expressions

- Boolean functions can also be represented as boolean (propositional) expressions.
- $x \wedge y$  represents the function:
  - $f : \{0, 1\} \times \{0, 1\} \rightarrow \{0, 1\}$ 
    - $f(0, 0) = 0$
    - $f(0, 1) = 0$
    - $f(1, 0) = 0$
    - $f(1, 1) = 1$

# Boolean functions and expressions

<b>x</b>	<b>y</b>	<b>z</b>	<b>g</b>
0	0	0	0
0	0	1	1
0	1	0	1
0	1	1	0
1	0	0	1
1	0	1	0
1	1	0	0
1	1	1	1

$$g : \{0, 1\} \times \{0, 1\} \times \{0, 1\} \rightarrow \{0, 1\}$$

$$g = ((x \Leftrightarrow y) \wedge z) \vee ((x \Leftrightarrow \neg y) \wedge \neg z)$$

# Boolean expressions and functions

<b>x</b>	<b>y</b>	<b>z</b>	<b>g</b>
0	0	0	
0	0	1	
0	1	0	
0	1	1	
1	0	0	
1	0	1	
1	1	0	
1	1	1	

$$g = (x \wedge y \wedge \neg z) \vee (x \wedge \neg y \wedge z) \vee (\neg x \wedge y)$$

# Boolean expressions and functions

<b>x</b>	<b>y</b>	<b>z</b>	<b>g</b>
0	0	0	0
0	0	1	0
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1	0	1	1
1	1	0	1
1	1	1	0

$$g = (x \wedge y \wedge \neg z) \vee (x \wedge \neg y \wedge z) \vee (\neg x \wedge y)$$

$$g : \{0, 1\} \times \{0, 1\} \times \{0, 1\} \rightarrow \{0, 1\}$$



# Three Representations

- *Boolean functions*
- *Truth tables*
- *Propositional formulas.*
- Three *equivalent* representations.
- Here is a *fourth one!*

# Boolean Decision Tree

- A *boolean function* is represented as a *(binary) tree*.
- Each *internal node* is labeled with a (boolean) *variable*.
- Each *internal node* has a *positive (full line)* and a *negative (dotted line) successor*.
- The *terminal nodes* are labeled with **0** or **1**.

# Boolean Decision Diagrams

- A **compact way** of representing boolean functions.
- Can be used in **CTL** model checking.
  - Represent a subset of states as a boolean function.
  - Represent the transition relation as a boolean function.
  - Reduce **EX( $\psi$ )**, **EU( $\psi_1, \psi_2$ )** and **EG( $\psi$ )** to manipulating boolean functions and checking for **boolean function equality**.
- Go from **NuSMV** (program) representation *directly* to its **BDD** representation!

# If-Then-Else operator

$$(x \rightarrow s_1, s_0) \equiv (x \wedge s_1) \vee (\neg x \wedge s_0)$$

x	y	z	$x \rightarrow y, z$
0	0	0	0
0	0	1	1
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	0
1	1	0	1
1	1	1	1

x	y	$x \rightarrow y, 0$	$x \wedge y$
0	0	0	0
0	1	0	0
1	0	0	0
1	1	1	1

x	y	$x \rightarrow 1, y$	$x \vee y$
0	0	0	0
0	1	1	1
1	0	1	1
1	1	1	1

# If-Then-Else representation

Let  $x \in AP$ , then

- $x \equiv x \rightarrow 1, 0$
- $\neg\varphi \equiv \varphi \rightarrow 0, 1$
- $\varphi_1 \wedge \varphi_2 \equiv \varphi_1 \rightarrow \varphi_2, 0$
- $\varphi_1 \vee \varphi_2 \equiv \varphi_1 \rightarrow 1, \varphi_2$

***Theorem:*** Every boolean formula can be written in *If-Then-Else representation*.

Assume  $\varphi_1 \equiv x \rightarrow \psi_1, \psi_2$  then

$$\begin{aligned} \varphi_1 \rightarrow \varphi_2, \varphi_3 &\equiv (x \rightarrow \psi_1, \psi_2) \rightarrow \varphi_2, \varphi_3 \equiv \\ &\equiv x \rightarrow (\psi_1 \rightarrow \varphi_2, \varphi_3), (\psi_2 \rightarrow \varphi_2, \varphi_3) \end{aligned}$$

# If-Then-Else normal form

***ITE normal form***: a boolean expression is written in *ITE normal form* if it only contains constants 0 and 1, If-Then-Else is the only operator occurring in the expression and tests are only performed on variables.

# Boolean decision trees.

## *If-Then-Else normal form*

$$\mathbf{x} \wedge \mathbf{y} = \mathbf{x} \rightarrow \mathbf{y}, \mathbf{0}$$

## *Shannon Expansion:*

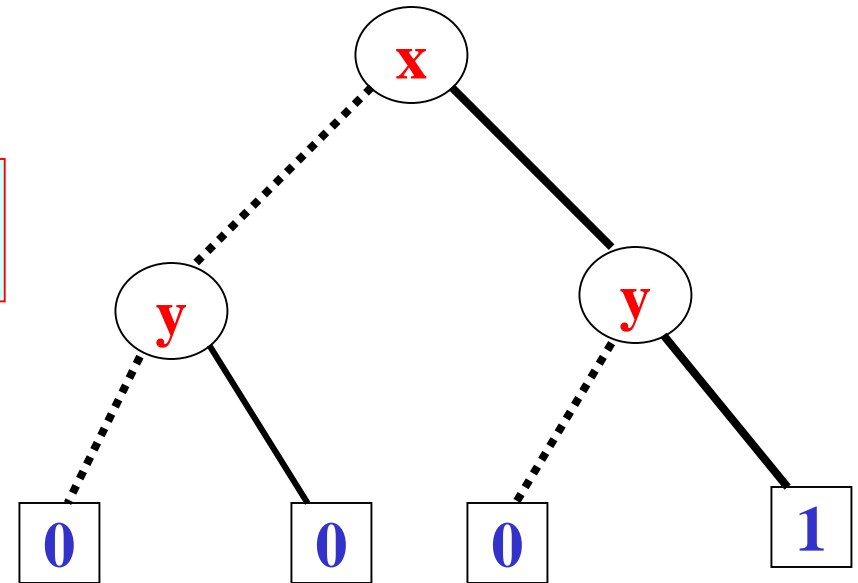
$$\mathbf{f} = (\mathbf{x} \wedge \mathbf{f}_{[1/\mathbf{x}]}) \vee (\neg \mathbf{x} \wedge \mathbf{f}_{[0/\mathbf{x}]})$$

$$\mathbf{f} = \mathbf{x} \rightarrow \mathbf{f}_{[1/\mathbf{x}]}, \mathbf{f}_{[0/\mathbf{x}]}$$

where

$$\mathbf{f}_{[a/\mathbf{x}]}(\dots, \mathbf{x}, \dots) = \mathbf{f}(\dots, \mathbf{a}, \dots)$$

for  $\mathbf{a} = \mathbf{0}, \mathbf{1}$ .



$$\mathbf{x} \wedge \mathbf{y}$$

# If-Then-Else normal form

***ITE normal form***: a boolean expression is written in *ITE normal form* if it only contains constants 0 and 1, If-Then-Else is the **only** operator occurring in the expression and **tests** are **only** performed on variables.

***Theorem***: Every boolean formula can be written in *ITE normal form*.

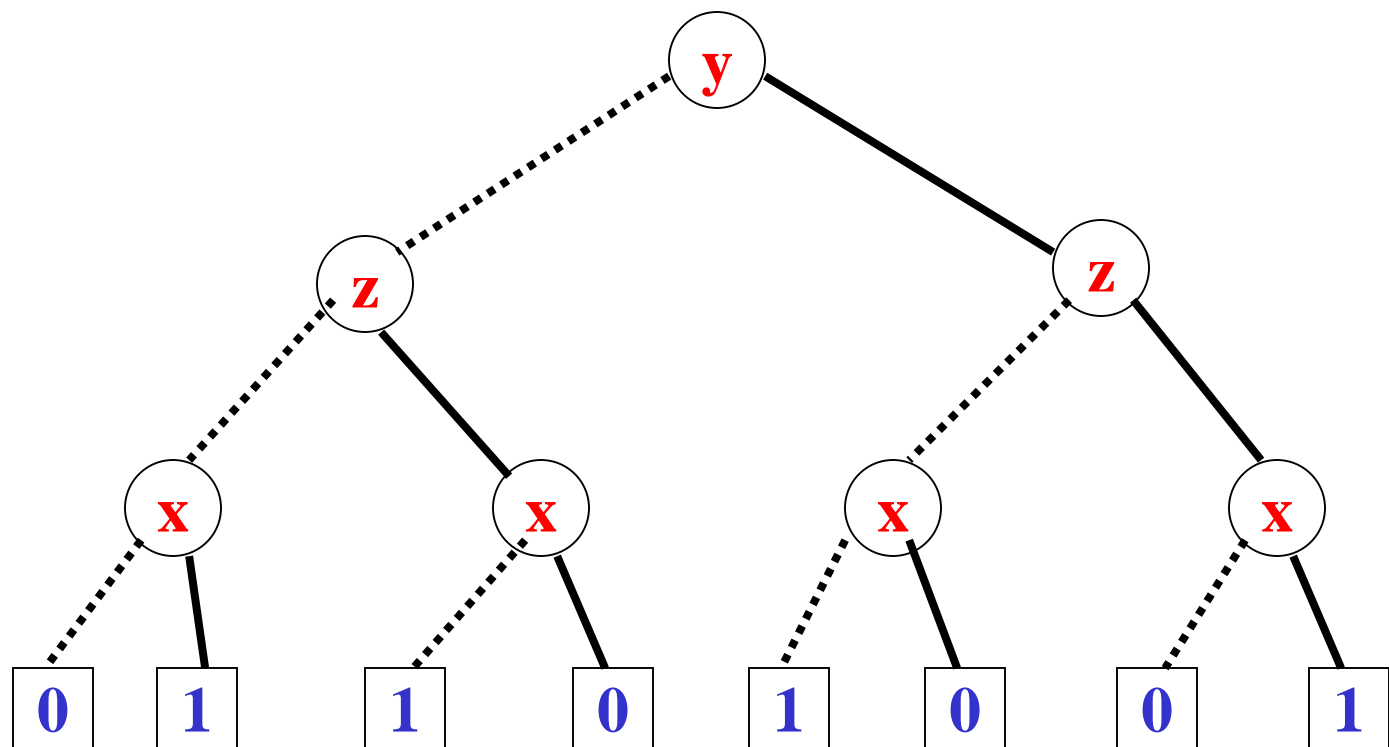
**Proof**: by trivial induction on the structure of boolean formulae.



# Boolean Decision Tree

- A *boolean function* is represented as a *(binary) tree*.
- Each *node* is *labeled* with a (boolean) *variable*.
- Each *node* has a *positive (full line)* and a *negative (dotted line) successor*.
- The *terminal nodes* are labeled with **0** or **1**.

<b>x</b>	<b>y</b>	<b>z</b>	<b>g</b>
0	0	0	0
0	0	1	1
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0	1	1	0
1	0	0	1
1	0	1	0
1	1	0	0
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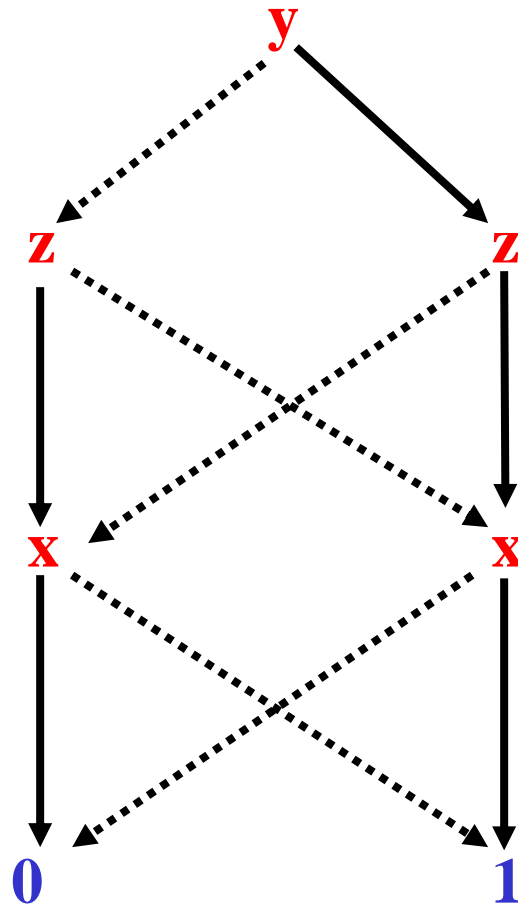


$$g = (y \wedge (x \Leftrightarrow z)) \vee (\neg y \wedge (x \Leftrightarrow \neg z))$$

# BDDs

A **BDD** is *finite rooted directed acyclic graph* in which:

- There is a *unique initial node* (the *root*)
- Each *terminal node* is labeled with a **0** or **1**.
- Each *non-terminal* (internal) node  $v$  has three attribute:
  - *var(v)*, and
  - exactly *two successors low(v)* and *high(v)*: one labeled **0** (*dotted edge, low(v)*) and the other labeled **1** (*solid edge, high(v)*).



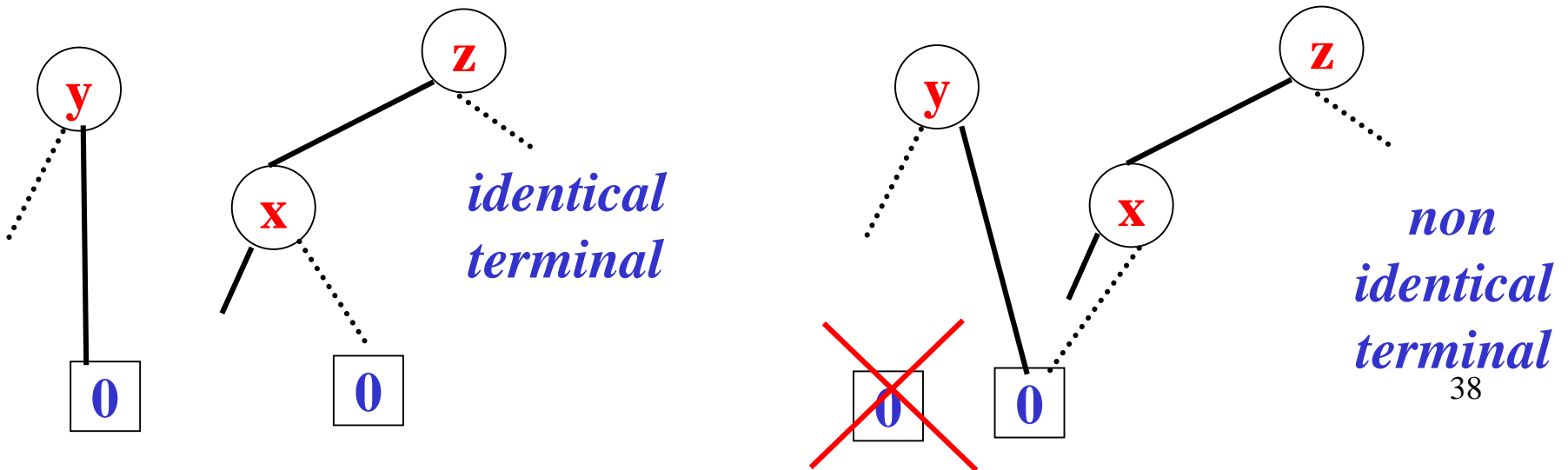
$$g = (y \wedge (x \Leftrightarrow z)) \vee (\neg y \wedge (x \Leftrightarrow \neg z))$$

# Reduction Rules

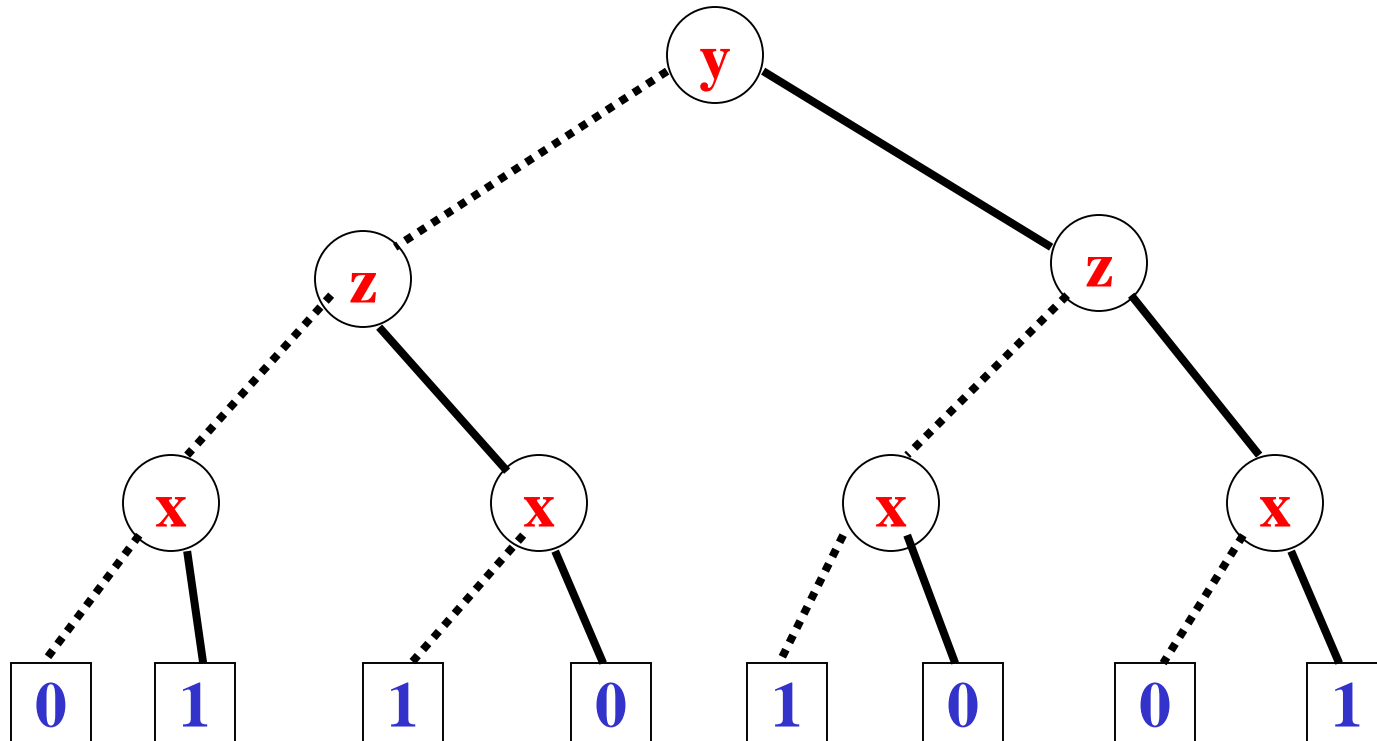
- Three reduction rules:
  - **Share identical terminal nodes. (R1)**
  - **Remove redundant tests (R2)**
  - **Share identical non-terminal nodes. (R3)**

# Reduction Rules

- Three reduction rules:
  - **Share identical terminal nodes. (R1)**
- If a **BDD** contains *two terminal nodes* **m** and **n** both *labeled 0* then, *remove n* and *direct all incoming edges at n to m*.
- **Similarly for two terminal nodes labeled 1.**

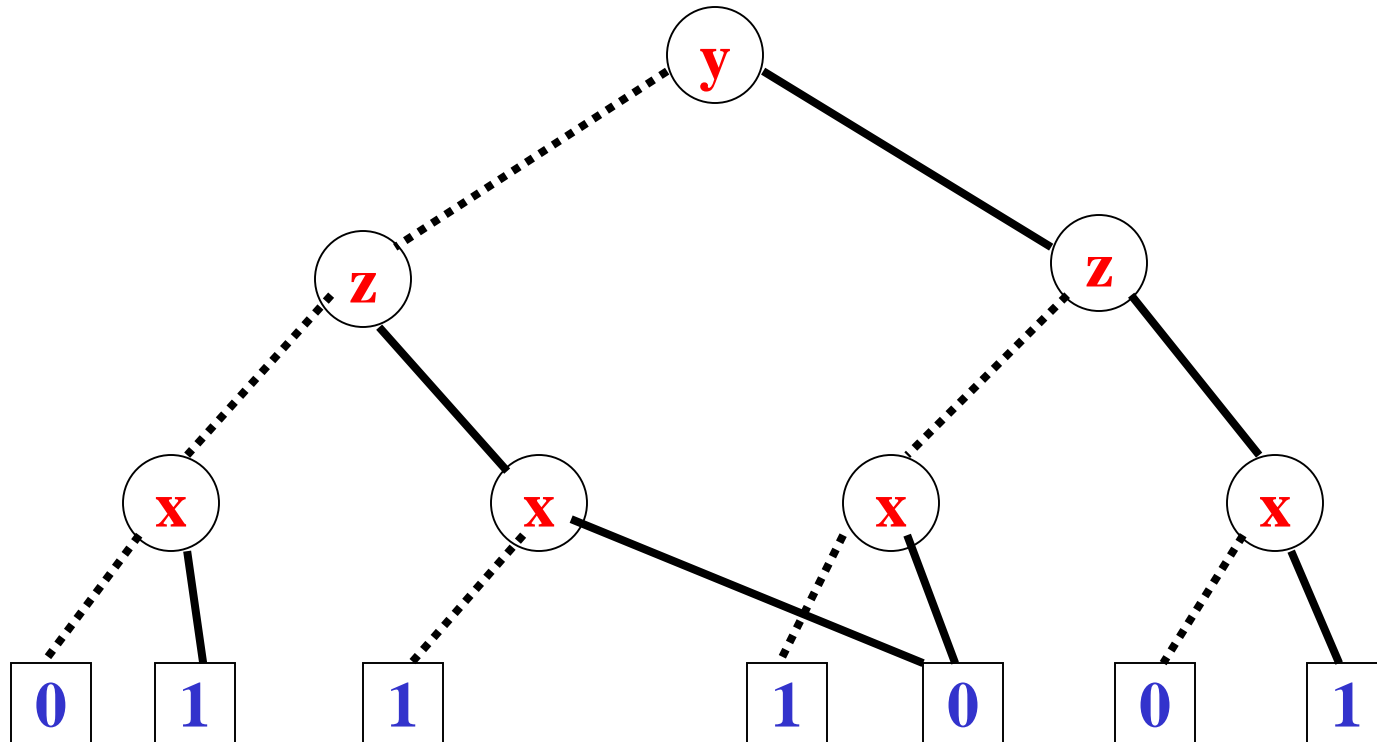


# Share identical terminal nodes. (R1)



$$g = (y \wedge (x \leftrightarrow z)) \vee (\neg y \wedge (x \leftrightarrow \neg z))$$

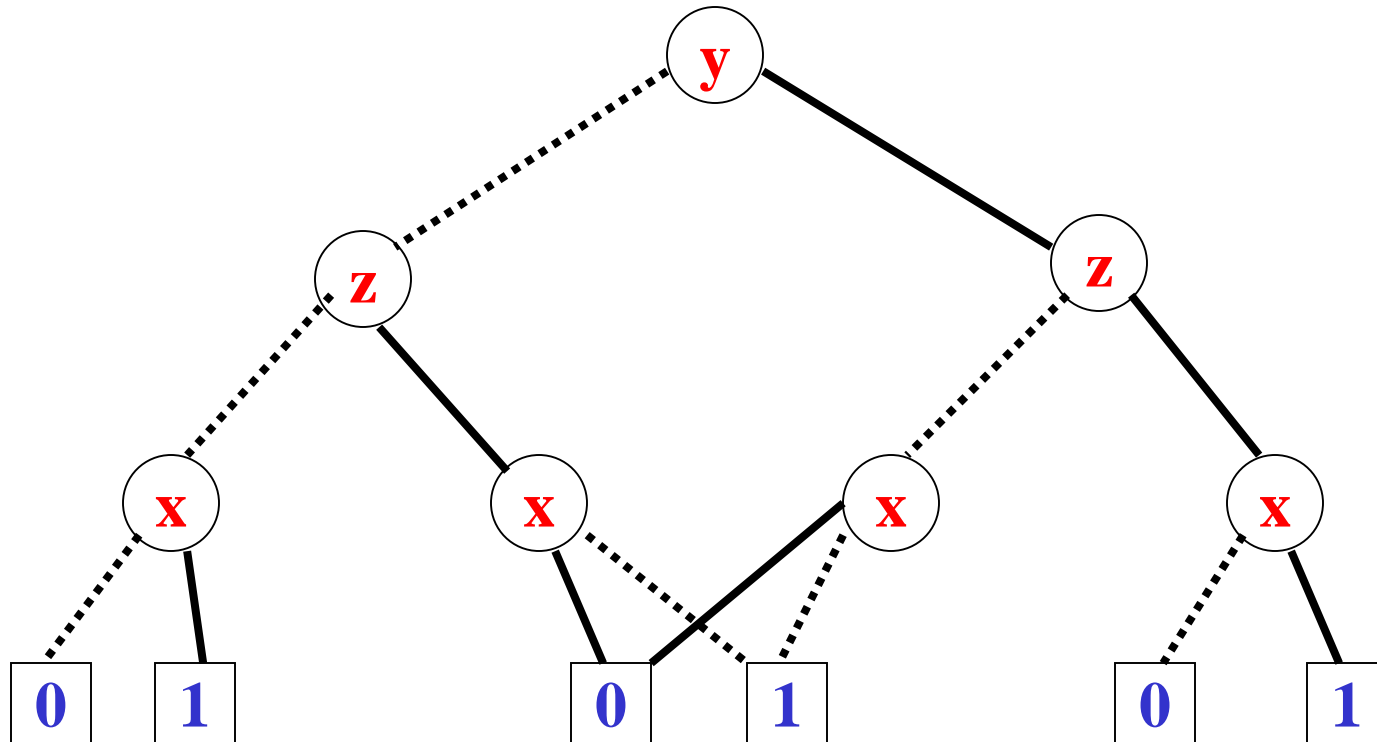
# Share identical terminal nodes. (R1)



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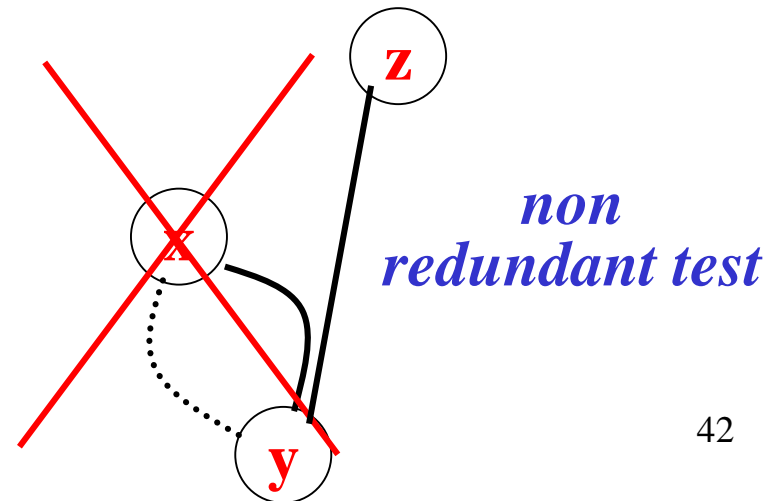
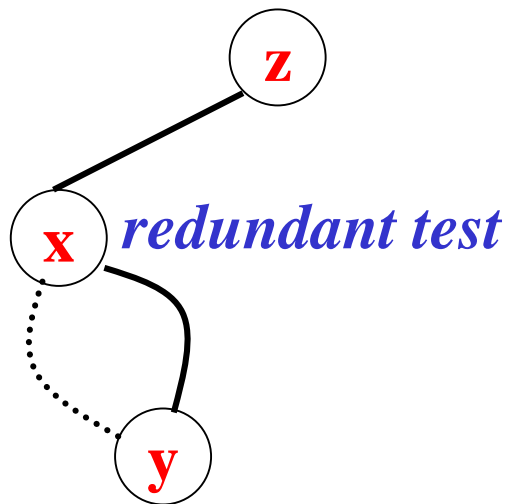
# Share identical terminal nodes. (R1)



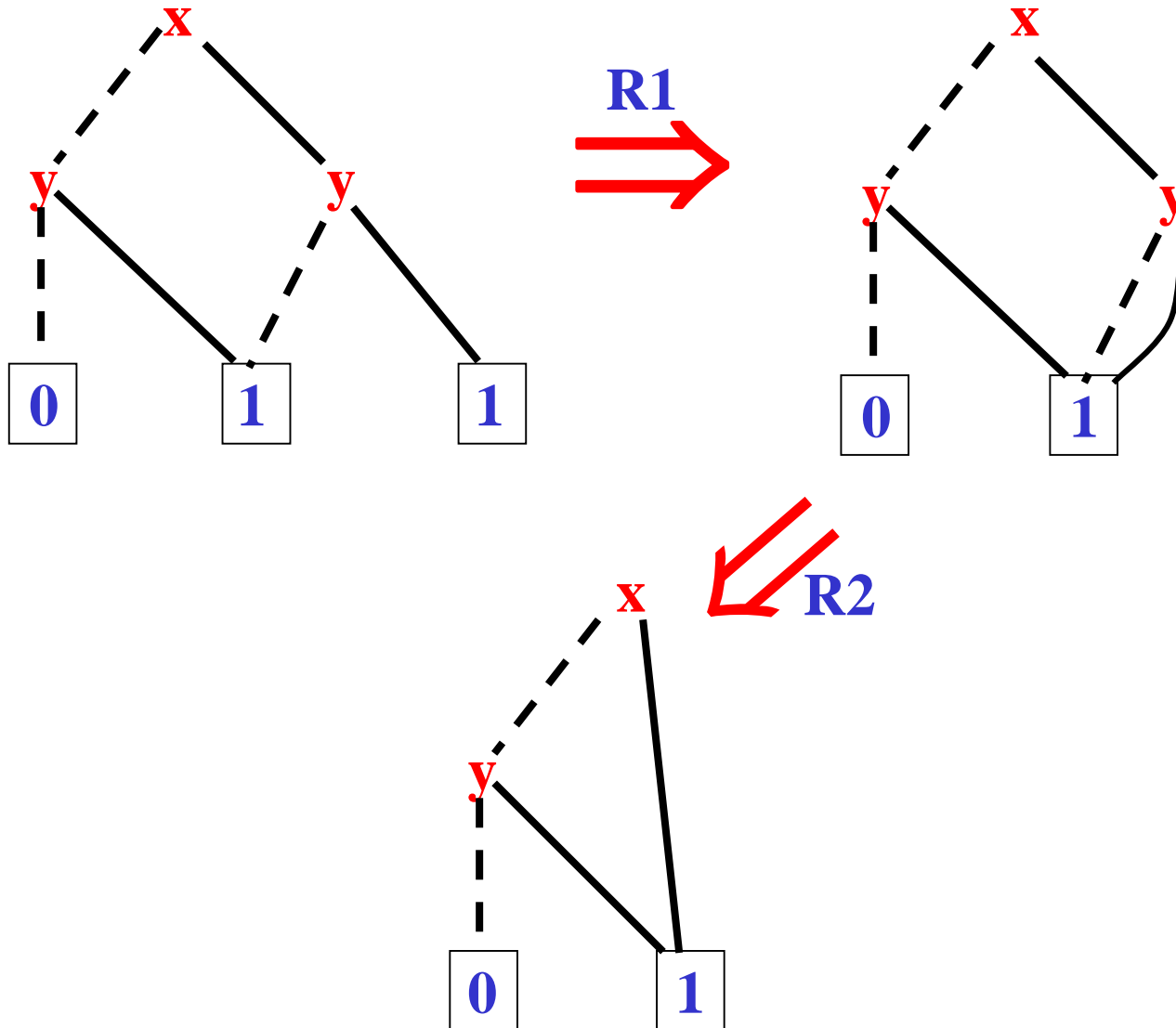
$$g = (y \wedge (x \leftrightarrow z)) \vee (\neg y \wedge (x \leftrightarrow \neg z))$$

# Reduction Rules

- Three reduction rules:
  - Share identical terminal nodes. (**R1**)
  - **Remove redundant tests (R2)**
- If both *successors of node m lead to the same node n* then *remove m* and *direct all incoming edges of m to n*.

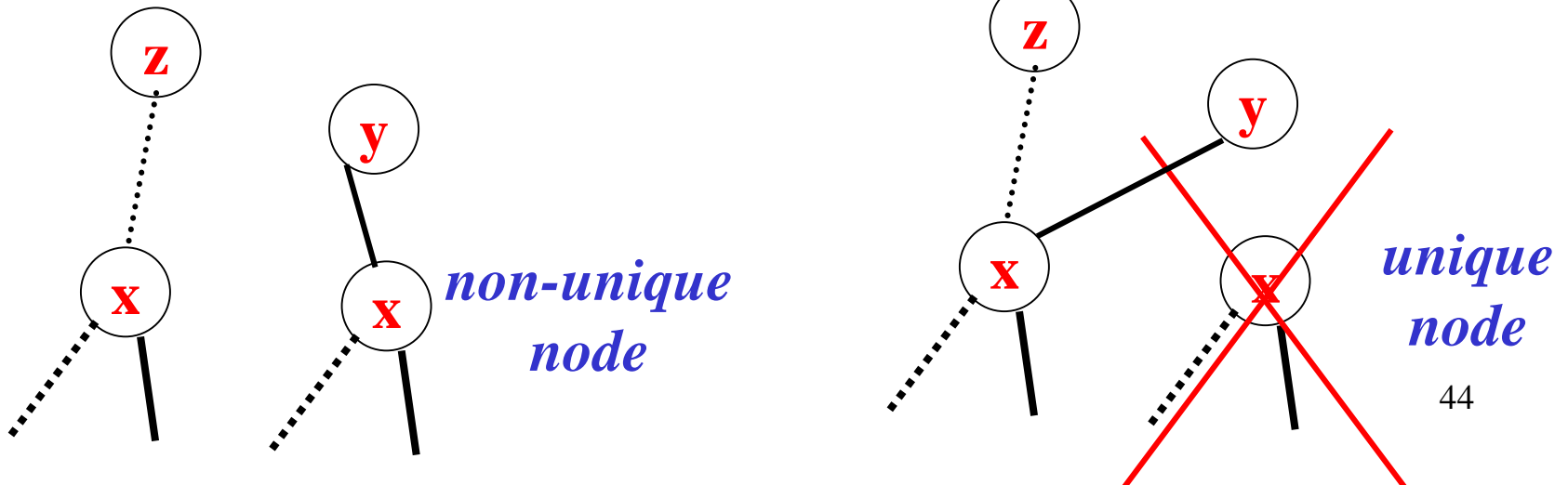


# Remove redundant tests (R2)

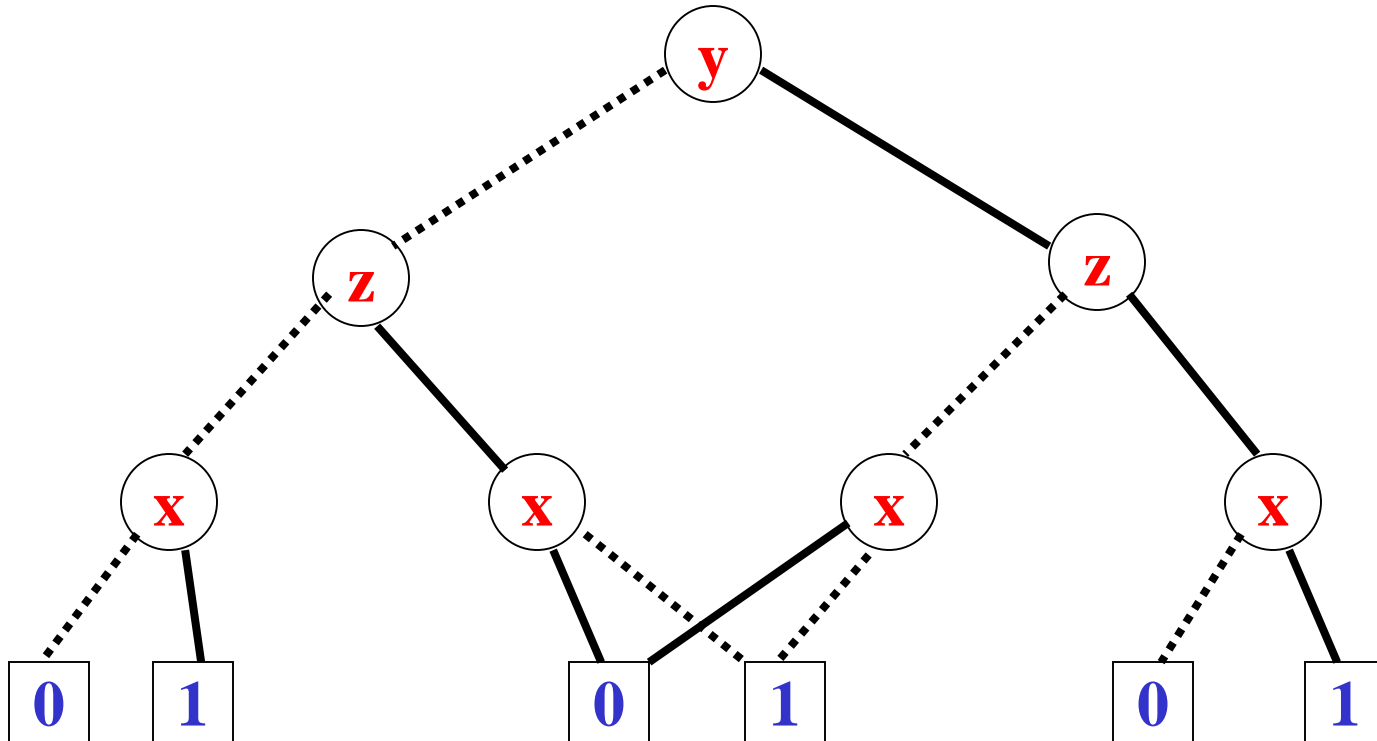


# Reduction Rules

- Three reduction rules:
  - Share identical terminal nodes. (**R1**)
  - Remove redundant tests (**R2**)
  - **Share identical non-terminal nodes. (R3)**
- If the *sub-BDDs rooted at the nodes m* and *n* are “*identical*” then *remove m* and *direct all its incoming edges to n*.

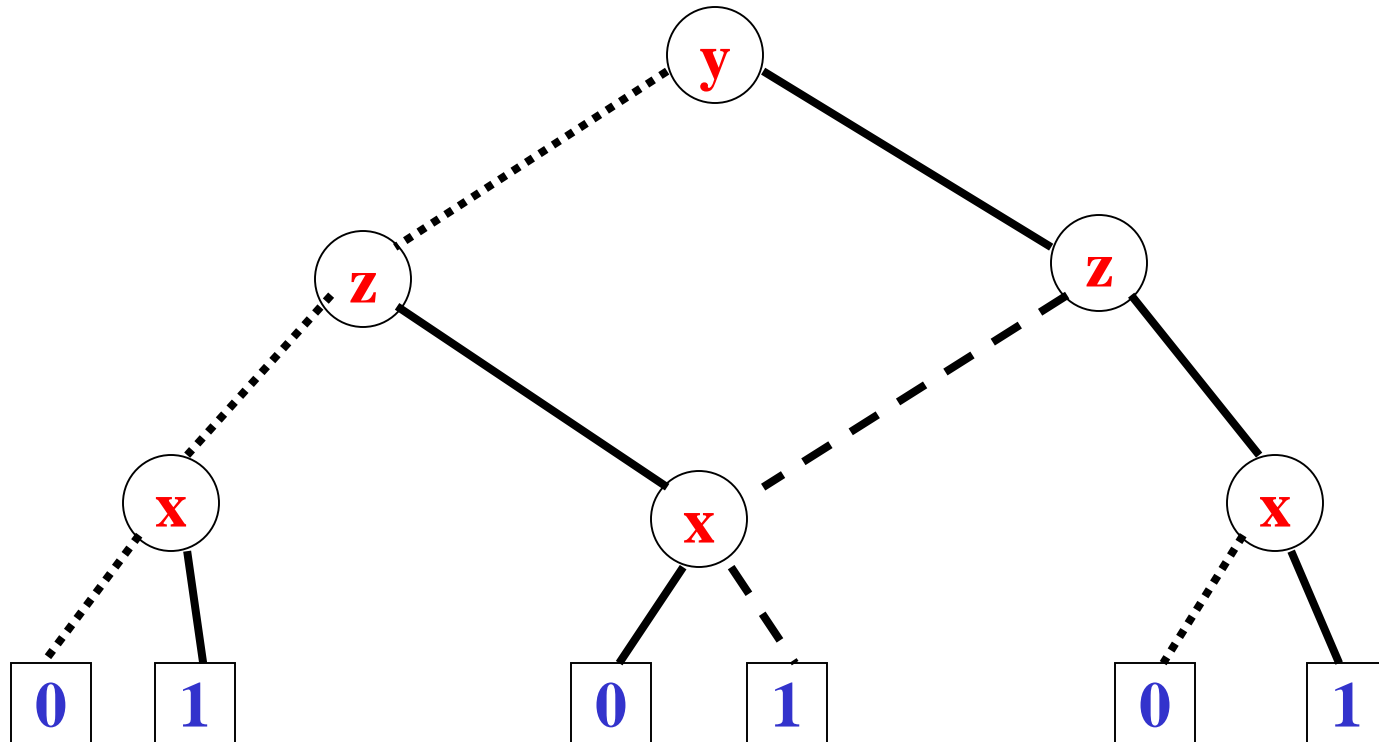


# Share identical non-terminal nodes. (R3)



$$g = (y \wedge (x \leftrightarrow z)) \vee (\neg y \wedge (x \leftrightarrow \neg z))$$

# Share identical non-terminal nodes. (R3)



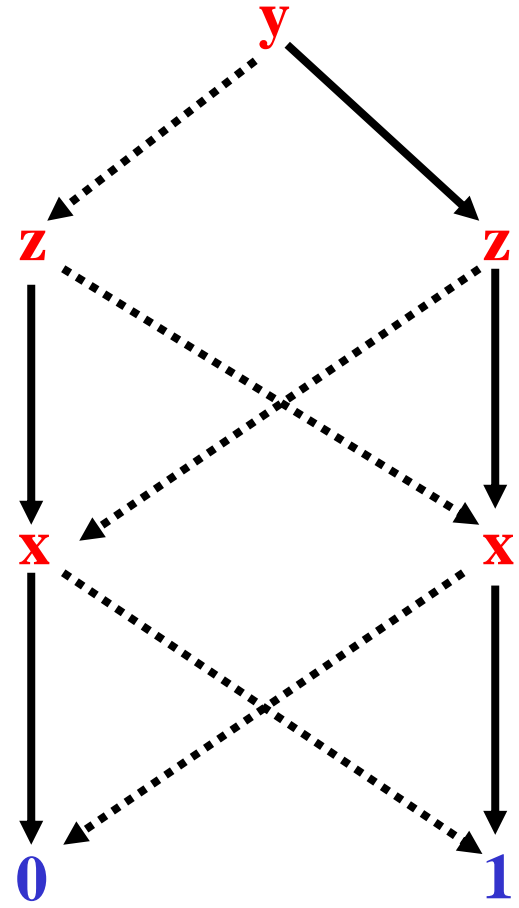
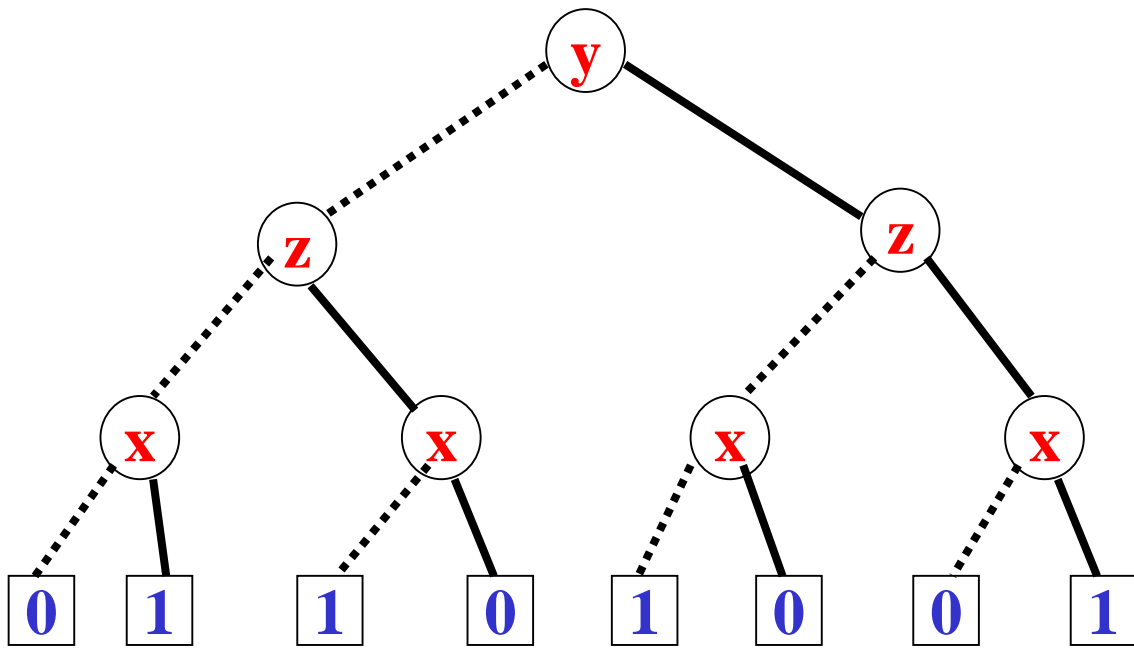
$$g = (y \wedge (x \leftrightarrow z)) \vee (\neg y \wedge (x \leftrightarrow \neg z))$$

# Reduced BDDs

- A **BDD** is *reduced iff* none of the three reduction rules can be applied to it.
- Start from the bottom layer (terminal nodes).
- *Apply* the *rules* repeatedly *to level i*. And then *move to level i-1* (in this way checking for applicability of R3 only needs testing whether **var(m)=var(n)**, **low(m)=low(n)** and **high(m)=high(n)**).
- Stop when the root node has been treated.
- This can be done efficiently.

# Binary Decision Tree for

# Reduced BDD



$$g = (y \wedge (x \Leftrightarrow z)) \vee (\neg y \wedge (x \Leftrightarrow \neg z))$$



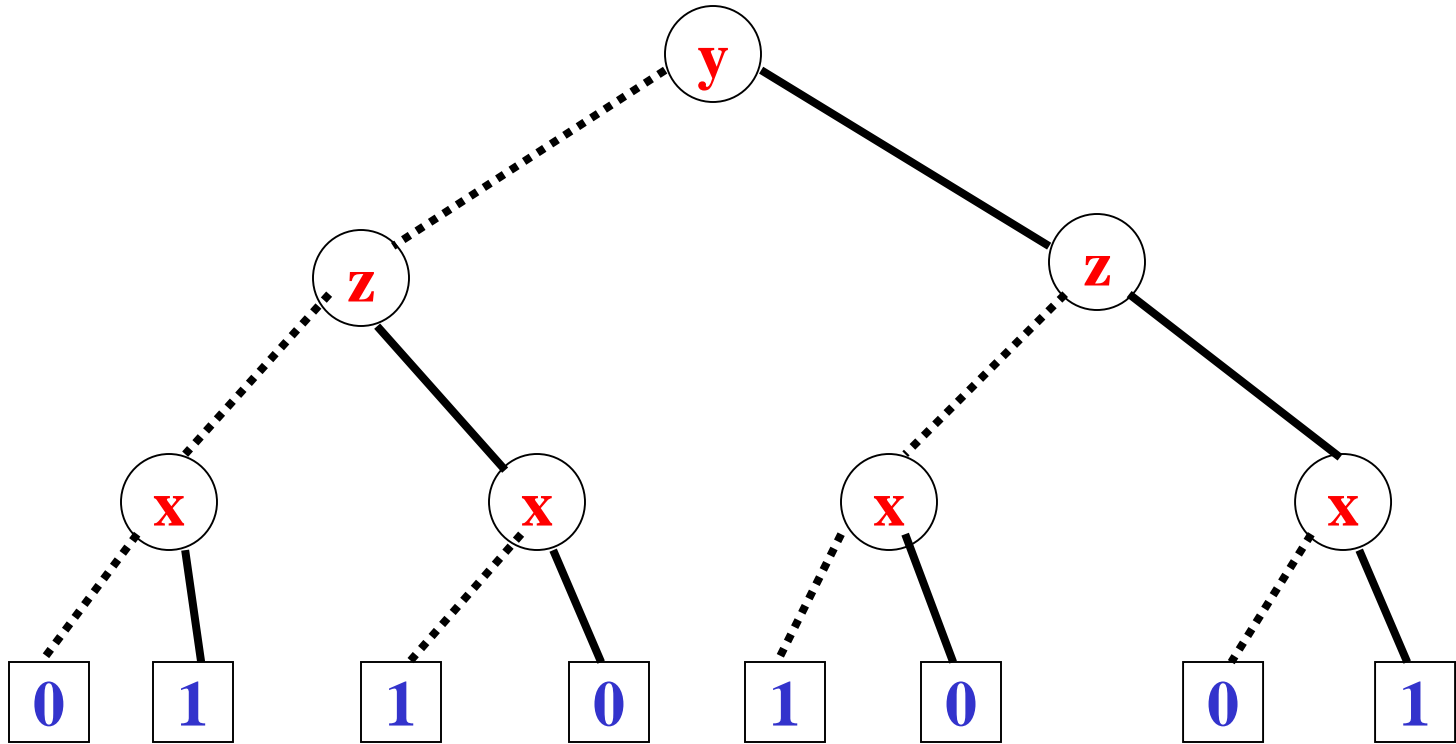
# Ordered BDDs

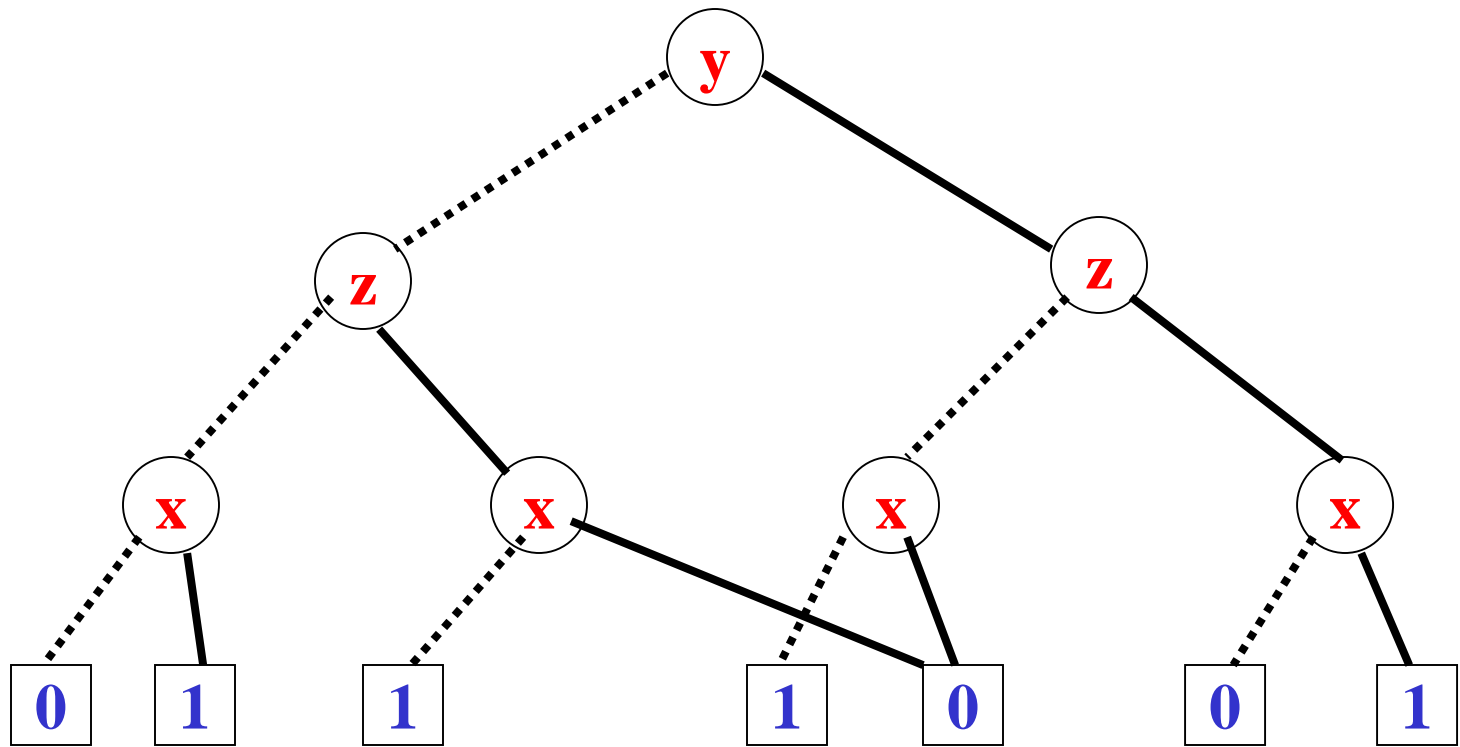
- $\{x_1, x_2, \dots, x_n\}$ 
  - An indexed (ordered) set of boolean variables.
  - $x_1 < x_2 \dots < x_n$
- **G** is an **ordered BDD** w.r.t. the above *variable ordering iff*:
  - Each variable that appears in **G** is in the above set. (but the converse may not be true).
  - If  $i < j$  and  $x_i$  and  $x_j$  appear on a path then  $x_i$  **appears before  $x_j$** .

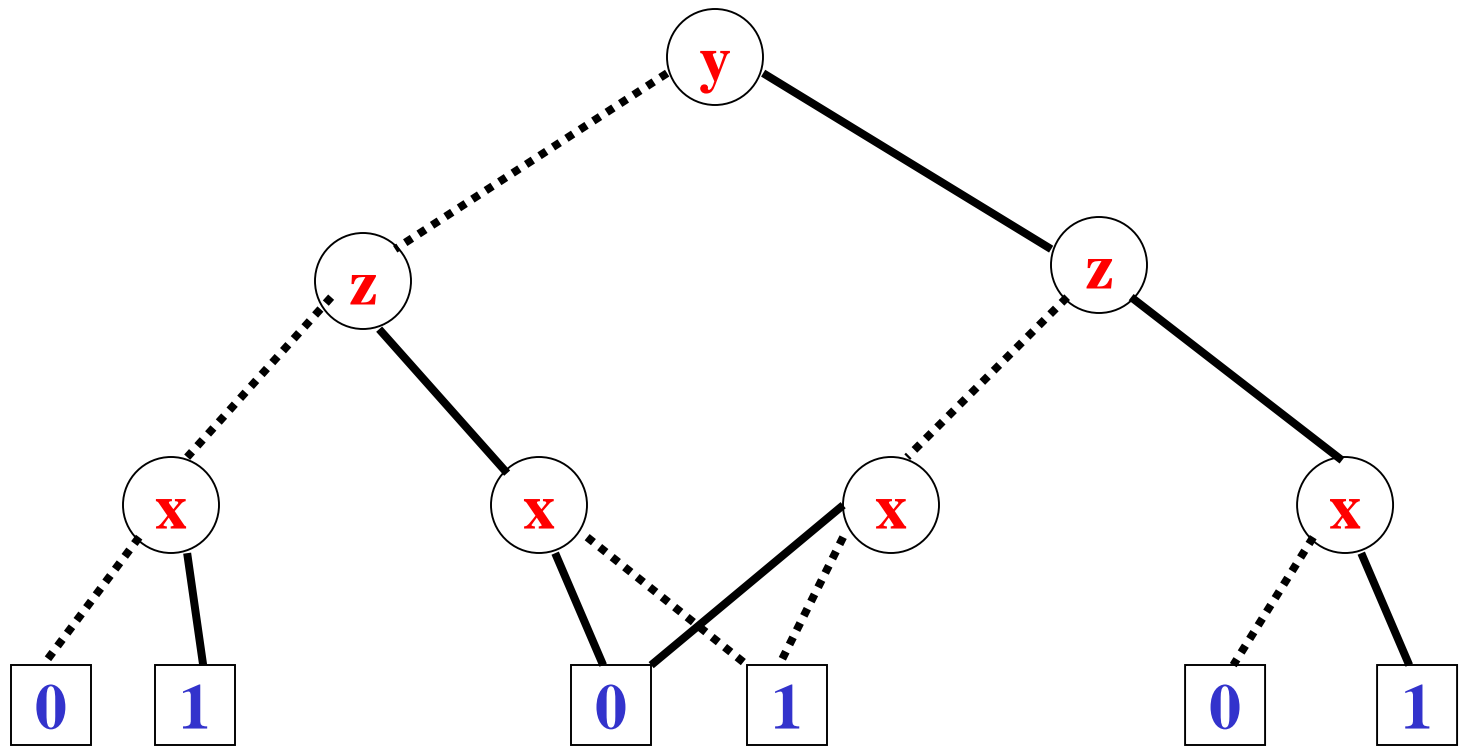
# Ordered BDDs

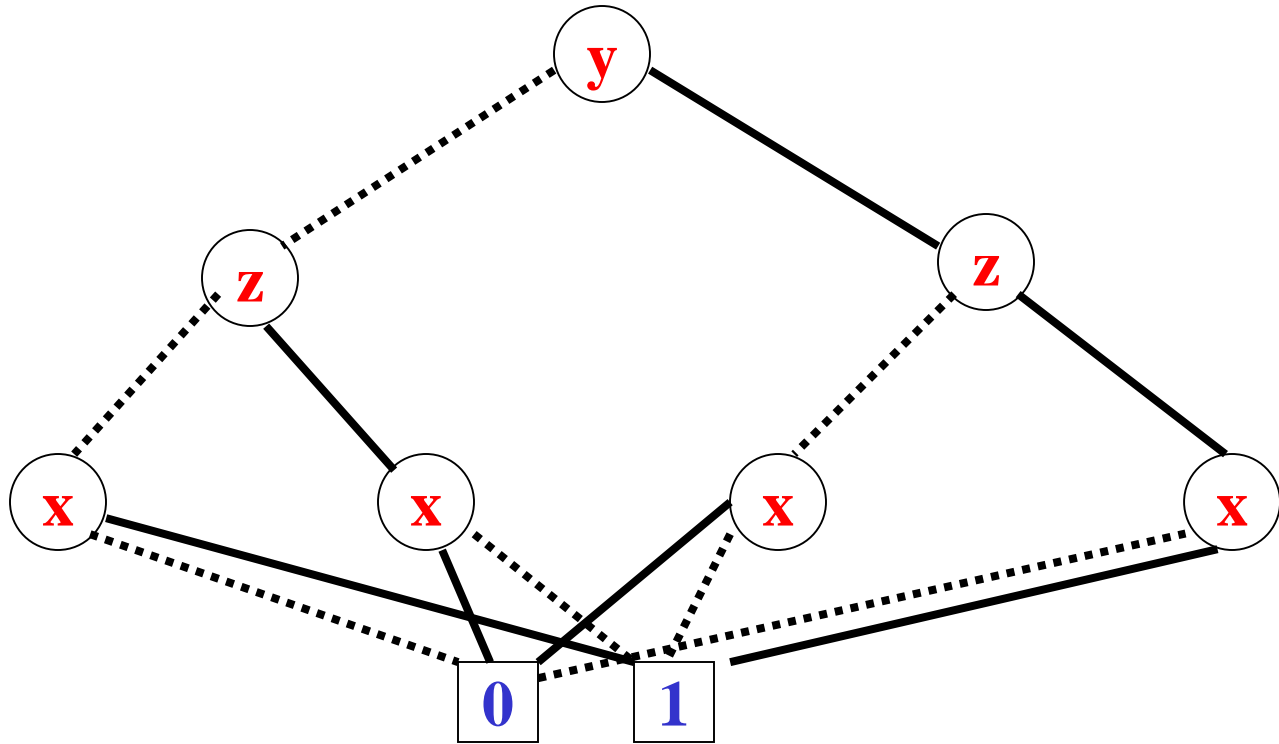
- Fundamental Fact:
  - For a fixed variable ordering, each boolean function has *exactly one* reduced Ordered BDD!
  - Reduced OBDDs are *canonical objects*.
  - To test if  $f$  and  $g$  are equal, we just have to check if **their** reduced **OBDDs** are **identical**.
  - This will be crucial for model checking!

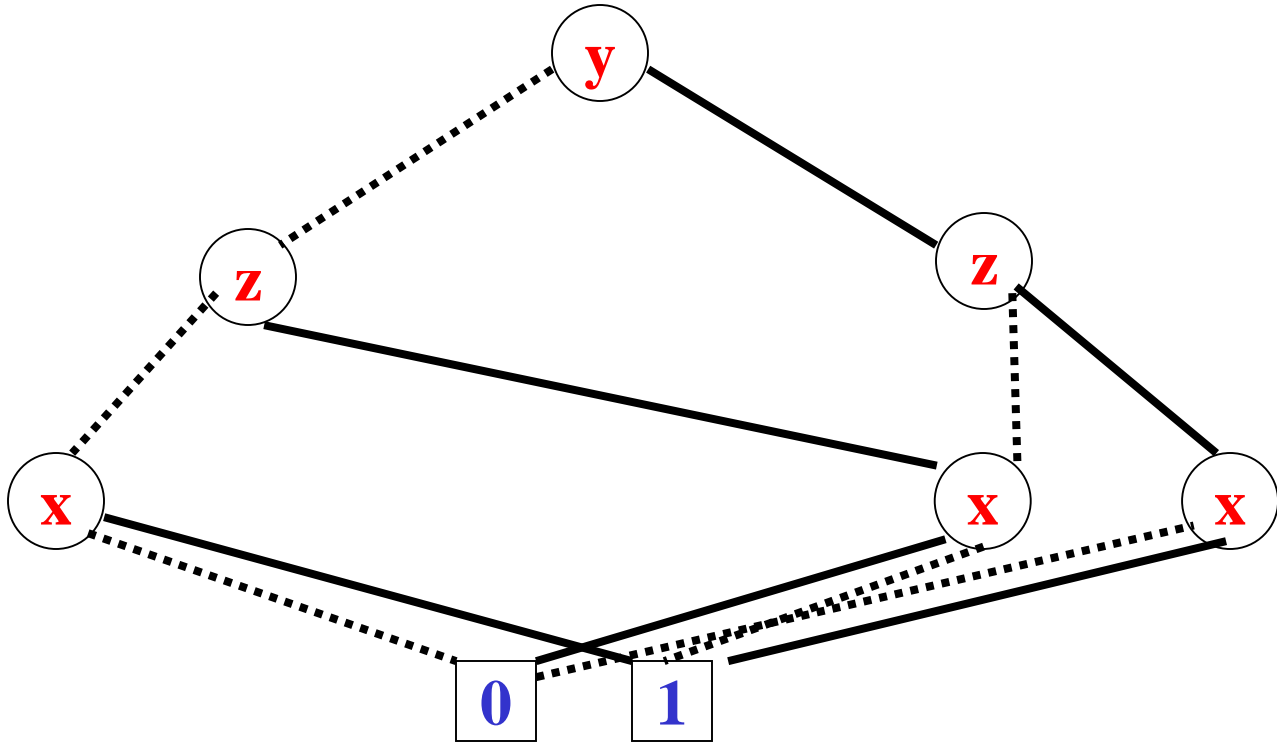
$y < z < x$

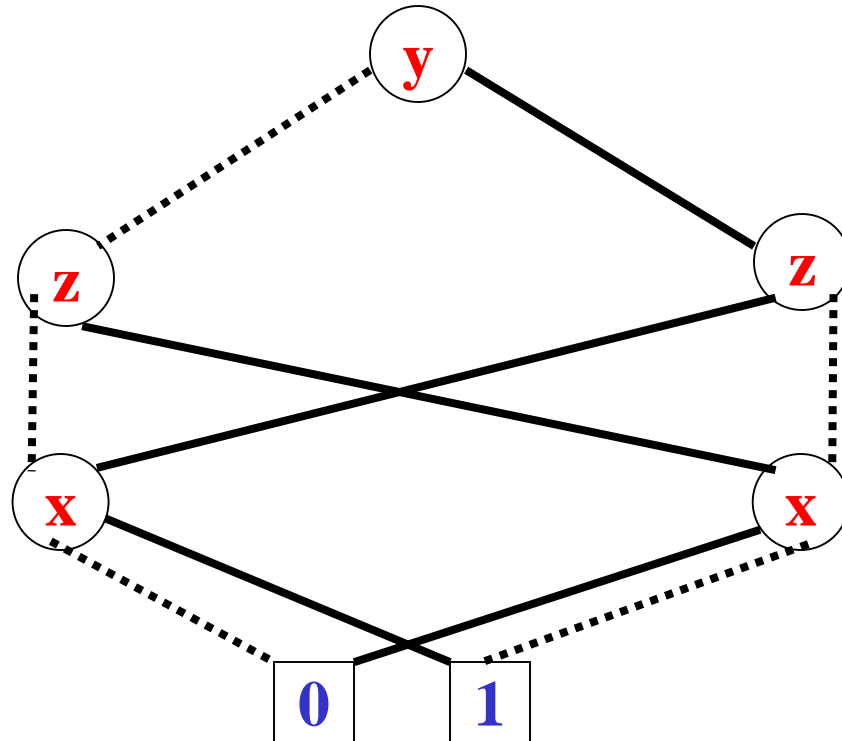














# Reduced OBDD

- An **OBDD** is *reduced* (i.e. it is a **ROBDD**) if there are only *two terminal vertices* **0** and **1**, and for all *non terminal vertices*  $v, u$ :
  - $low(v) \neq high(v)$  (*non-redundant tests*)
  - $low(v) = low(u)$ ,  $high(v) = high(u)$  and  $var(v) = var(u)$  implies  $v = u$  (*uniqueness*)

# Canonicity of ROBDD

Let us denote a **ROBDD** with its *root node* and the *function* represented by *subgraph a rooted* in node  $u$  with  $f^u$ . Then:

**Theorem:** For any function  $f:\{0,1\}^n \rightarrow \{0,1\}$  *there exists a unique ROBDD  $u$*  with variable ordering  $x_1, x_2, \dots, x_n$  such that

$$f^u = f(x_1, \dots, x_n)$$

# Consequences of canonicity

*Theorem*: For any function  $f:\{0,1\}^n \rightarrow \{0,1\}$  there exists a **unique ROBDD  $u$**  with variable ordering  $x_1, x_2, \dots, x_n$  such that

$$f^u = f(x_1, \dots, x_n)$$

Therefore we can say that:

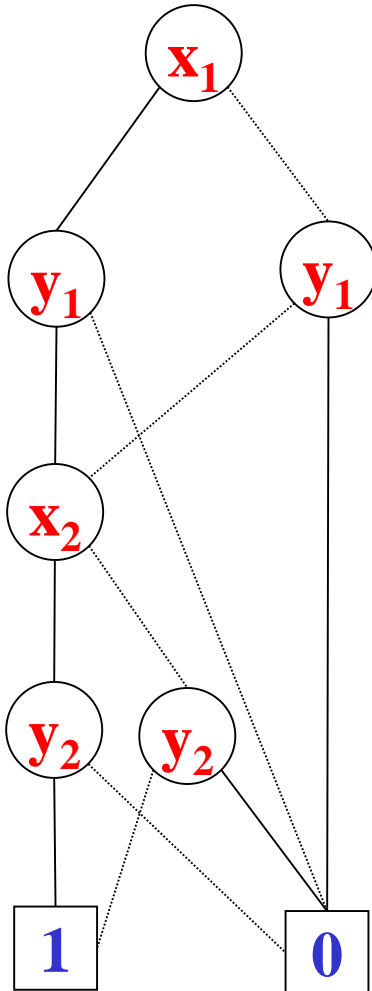
- A function  $f^u$  is a **tautology** if its **ROBDD  $u$**  is **equal** to **1**.
- A function  $f^u$  is a **satisfiable** if its **ROBDD  $u$**  is **not equal** to **0**.

# Reduced OBDDs

- *The ordering is crucial!*
- $\{x_1, x_2, y_1, y_2\}$       $x_1 \ x_2$ 
  - $f(x_1, x_2, y_1, y_2)$       $y_1 \ y_2$
  - $f(x_1, x_2, y_1, y_2) = 1$  *iff*  $(x_1 = y_1 \wedge x_2 = y_2)$
- If  $x_1 < y_1 < x_2 < y_2$ , then the **OBDD** is of size  $3 \cdot 2 + 2 = 8$ .
- If  $x_1 < x_2 < y_1 < y_2$ , then the **OBDD** is of size  $3 \cdot 2^2 - 1 = 11$  !

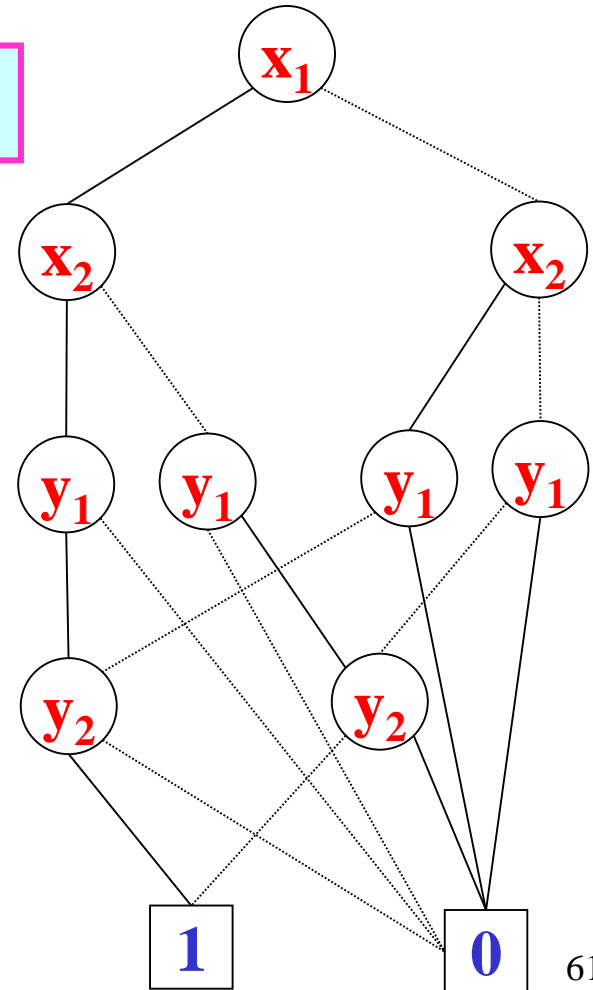
# Reduced OBDDs

$$x_1 < y_1 < x_2 < y_2$$



$$(x_1 = y_1 \wedge x_2 = y_2)$$

$$x_1 < x_2 < y_1 < y_2$$



# Reduced OBDDs

- *The ordering is crucial!*

- $\{x_1, x_2, \dots, x_n, y_1, y_2, \dots, y_n\}$        $x_1 \ x_2 \ \dots \ x_n$

$$f(x_1, x_2, \dots, x_n, y_1, y_2, \dots, y_n) \quad y_1 \ y_2 \ \dots \ y_n$$

$$- f(x_1, x_2, \dots, x_n, y_1, y_2, \dots, y_n) = 1 \quad \text{iff} \quad \bigwedge_{i=1}^n (x_i = y_i)$$

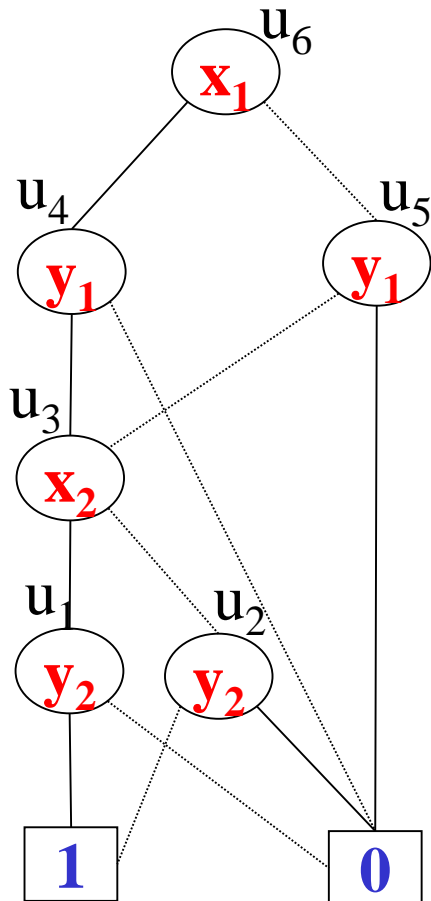
- If  $x_1 < y_1 < x_2 < y_2 \dots < x_n < y_n$ , then the **OBDD** is of size  $3n + 2$ .
- If  $x_1 < x_2 < \dots < x_n < y_1 < \dots < y_n$ , then the **OBDD** is of size  $3 \cdot 2^n - 1$  !

# ROBDDs

- Finding the *optimal variable ordering* is *computationally expensive* (**NP-complete**).
- There are *heuristics* for finding “*good orderings*”.
- There exist boolean functions whose sizes are *exponential* (in the number of variables) for any ordering.
- Functions encountered in practice are **rarely** of this kind.

# Implementation of ROBDDs

## *Array-based implementation*



$T[] =$

*root* =  $u_6$

	Var	Low	High
<b>0</b>	?	?	?
<b>1</b>	?	?	?
<b><math>u_1</math></b>	<b><math>y_2</math></b>	<b>0</b>	<b>1</b>
<b><math>u_2</math></b>	<b><math>y_2</math></b>	<b>1</b>	<b>0</b>
<b><math>u_3</math></b>	<b><math>x_2</math></b>	<b><math>u_2</math></b>	<b><math>u_1</math></b>
<b><math>u_4</math></b>	<b><math>y_2</math></b>	<b>0</b>	<b><math>u_3</math></b>
<b><math>u_5</math></b>	<b><math>y_1</math></b>	<b>0</b>	<b><math>u_3</math></b>
<b><math>u_6</math></b>	<b><math>x_1</math></b>	<b><math>u_5</math></b>	<b><math>u_4</math></b>



# The function MK

- The function **MK** searches for a node  $u$  with  $var(u)=x_i$ ,  $low(u)=l$  and  $high(u)=h$ . If the node does not exist, then creates the new node after inserting it. The running time is  $O(1)$ .

$H(i,l,h)$  is a hash function mapping a triple  $\langle i,l,h \rangle$  into a node index in  $T$ .

**Algorithm mk(i,l,h)**

if  $l=h$  then

return  $l$

else if  $T[H(i,l,h)] \neq \text{empty}$  then

return  $T[H(i,l,h)]$

else  $u = \text{add}(T,H(i,l,h),i,l,h)$

return  $u$

# Operations on ROBDDs.

- During model checking, boolean operations will have to be performed on **ROBDDs**.
- These operations can be implemented efficiently.
- $f \vee g$  -----  $G_f \text{ op}_{\vee} G_g = G_{f \vee g}$
- There is a procedure called **APPLY** to do this.

# Operations on ROBDDs

- When performing an operation on  $G$  and  $G'$  we assume their variable orderings are *compatible*.
- $X = X_G \cup X_{G'}$ ,
- There is an ordering  $<$  on  $X$  such that:
  - $<$  restricted to  $X_G$  is  $<_G$
  - $<$  restricted to  $X_{G'}$  is  $<_{G'}$ .

# Operations on OBDDs

- The basic idea (Shannon Expansion):

- $f(x_1, x_2, \dots, x_n)$

- $f|_{x_1=0} = f(0, x_2, \dots, x_n)$

- $f = x_1 \vee (x_2 \wedge x_3)$

- $f|_{x_1=0} = x_2 \wedge x_3$

- Similarly,  $f|_{x_1=1} = f(1, x_2, \dots, x_n)$

$$f(x_1, x_2, \dots, x_n) = (\neg x_1 \wedge f|_{x_1=0}) \vee (x_1 \wedge f|_{x_1=1})$$

- This is true even if  $x_1$  does not appear in  $f$  !

# Operations on OBDDs: Negation

- The basic idea (Shannon Expansion):

$$f(x_1, x_2, \dots, x_n) = (\neg x_1 \wedge f|_{x_1=0}) \vee (x_1 \wedge f|_{x_1=1})$$

- Therefore, assuming  $x_1 < x_2 < \dots < x_n$ ,

$$\begin{aligned} \neg f(x_1, x_2, \dots, x_n) &= \neg ((\neg x_1 \wedge f|_{x_1=0}) \vee (x_1 \wedge f|_{x_1=1})) \\ &= (\neg(\neg x_1 \wedge f|_{x_1=0}) \wedge \neg(x_1 \wedge f|_{x_1=1})) \\ &= ((x_1 \vee \neg f|_{x_1=0}) \wedge (\neg x_1 \vee \neg f|_{x_1=1})) \\ &= (x_1 \wedge \neg x_1) \vee (\neg x_1 \wedge \neg f|_{x_1=0}) \vee \\ &\quad \vee (x_1 \wedge \neg f|_{x_1=1}) \vee (\neg f|_{x_1=0} \wedge \neg f|_{x_1=1}) \\ &= (\neg x_1 \wedge \neg f|_{x_1=0}) \vee (x_1 \wedge \neg f|_{x_1=1}) \end{aligned}$$

# Operations on ROBDDs.

- Let  $x$  be the top variable of  $G_f$  and  $y$  the top variable of  $G_g$ .
- To compute  $G_{f \text{ op } g}$  we consider:

**CASE1:  $x = y$**

$$\begin{aligned} \blacksquare f \text{ op } g = & (\neg x \wedge (f|_{x=0} \text{ op } g|_{x=0}) \vee \\ & (x \wedge (f|_{x=1} \text{ op } g|_{x=1})) \end{aligned}$$

- We have to solve now two **smaller** problems!

# Operations on ROBDDs.

- Let  $x$  be the top variable of  $G_f$  and  $y$  the top variable of  $G_g$ .

- To compute  $G_{f \text{ op } g}$  we consider:

**CASE2:  $x < y$ .**

– Then  $x$  does not appear in  $G_g$  (why?).

$$- \mathbf{g|_{x=0} = g = g|_{x=1}}$$

$$\blacksquare \mathbf{f \text{ op } g = (\neg x \wedge (f|_{x=0} \text{ op } g) \vee (x \wedge (f|_{x=1} \text{ op } g))}$$

– We have to solve now two **smaller** problems!

**CASE2:  $x > y$**  is symmetric.

# Operations on ROBDDs.

- To compute  $G_{f \text{ op } g}$  we consider:

**Base (terminal) cases** depend upon **op**

Eg.: if **op** =  $\vee$  then  $\{0,0 \rightarrow \mathbf{0}; \mathbf{1}\}$

if **op** =  $\wedge$  then  $\{1,1 \rightarrow \mathbf{1}; \mathbf{0}\}$

....

Notice that  $\neg f(x_1, x_2, \dots, x_n) = f(x_1, x_2, \dots, x_n) \oplus \mathbf{1}$ , therefore *negation* can be implemented with *Apply*.



# Algorithm for Apply

**Algorithm** `Apply(op,u,v)`

**Function** `App(u,v)`

**if** `terminal_case(op,u,v)` **then** **return** `op(u,v)`

**else if** `var(u) = var(v)` **then**

`u = mk(var(u), App(op,low(u),low(v)),  
App(op,high(u),high(v)))`

**else if** `var(u) < var(v)` **then**

`u = mk(var(u),App(op,low(u), v), App(op,high(u),v))`

**else** `/* var(u) > var(v) */`

`u = mk(var(u),App(op,u,low(v)), App(op,u,high(v)))`

**return** `u`

**return** `App(u,v)`

If  $n$  = number of variables, then  
*running time* =  $O(2^n)$ . Why?

# Efficient algorithm for Apply

**Algorithm** `Apply(op,u,v)`

**init**( $G_{op}$ )

**Function** `App(u,v)`

**if**  $G_{op}(u,v) \neq \text{empty}$  **then return**  $G_{op}(u,v)$

**else if** `terminal_case(op,u,v)` **then return** `op(u,v)`

**else if** `var(u)=var(v)` **then**

$r = \text{mk}(\text{var}(u), \text{App}(\text{op}, \text{low}(u), \text{low}(v)),$   
 $\text{App}(\text{op}, \text{high}(u), \text{high}(v)))$

**else if** `var(u) < var(v)` **then**

$r = \text{mk}(\text{var}(u), \text{App}(\text{op}, \text{low}(u), v), \text{App}(\text{op}, \text{high}(u), v))$

**else** */\** `var(u) > var(v)` *\*/*

$r = \text{mk}(\text{var}(u), \text{App}(\text{op}, u, \text{low}(v)), \text{App}(\text{op}, u, \text{high}(v)))$

$G_{op}(u,v) = r$

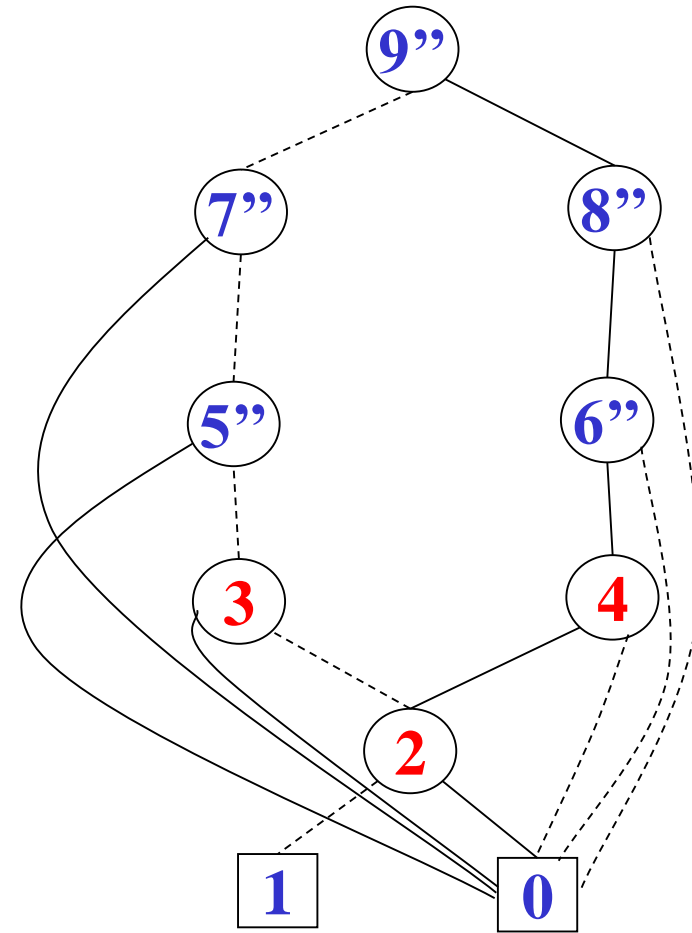
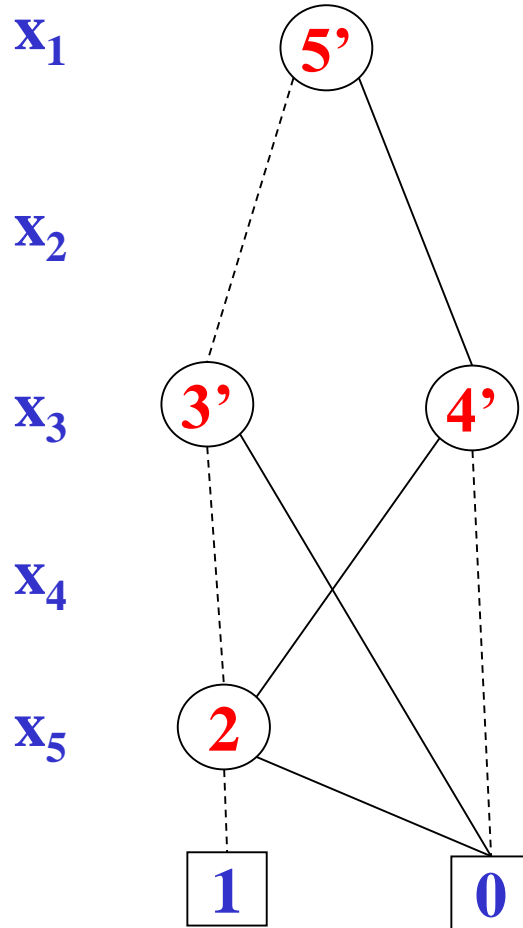
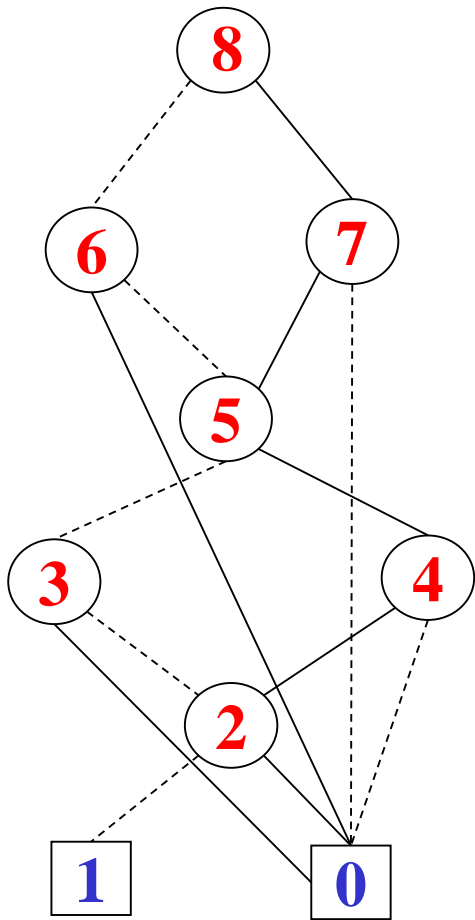
**return**  $r$

**return** `App(u,v)`

*running time* =  $O(|G_u| |G_v|)$ . Why?

# Example of Apply $\wedge$

$$\boxed{(x_1 \equiv x_2) \wedge (x_3 \equiv x_4) \wedge \neg x_5} \wedge \boxed{(x_1 \equiv x_3) \wedge \neg x_5} = \boxed{((x_1 \wedge x_2 \wedge x_3 \wedge x_4) \vee (\neg x_1 \wedge \neg x_2 \wedge \neg x_3 \wedge \neg x_4)) \wedge \neg x_5}$$



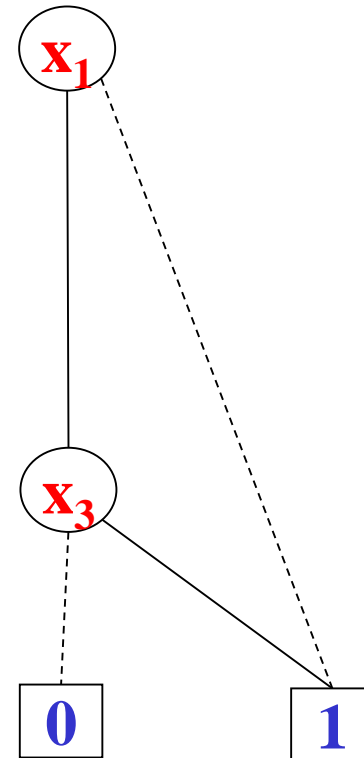
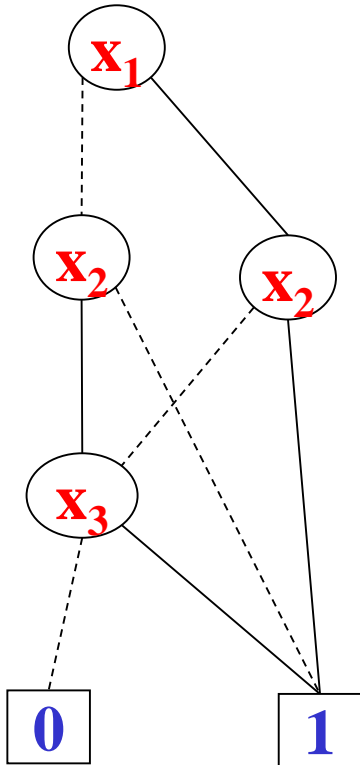
# The Restrict operation

- **Problem:** Given a (partial) truth assignment  $x_1=b_1, \dots, x_k=b_k$  (where  $b_j=0$  or  $b_j=1$ ), and a ROBDD  $t^u$ , compute the restriction of  $t^u$  under that assignment.
- E.G.: if  $f(x_1, x_2, x_3) = ((x_1 \Leftrightarrow x_2) \vee x_3)$  we want to compute  $f(x_1, x_2, x_3)[0/x_2] = f(x_1, 0, x_3)$   
i.e.:  $f(x_1, 0, x_3) = \neg x_1 \vee x_3$

# Restrict Operation: example

$$f(x_1, x_2, x_3) = ((x_1 \Leftrightarrow x_2) \vee x_3)$$

$$f(x_1, x_2, x_3)[0/x_2] = \neg x_1 \vee x_3$$



# Restrict Operation

- Let  $\mathbf{x}$  be the root of  $\mathbf{G}_f$
- To compute  $\mathbf{G}_f|_{\mathbf{y}=\mathbf{b}}$  we consider:

**CASE1:  $\mathbf{x} = \mathbf{y}$**

- $\mathbf{f}|_{\mathbf{y}=\mathbf{b}} = \mathbf{low}(\mathbf{G}_f)$       if  $\mathbf{b}=\mathbf{0}$
- $\mathbf{f}|_{\mathbf{y}=\mathbf{b}} = \mathbf{high}(\mathbf{G}_f)$       if  $\mathbf{b}=\mathbf{1}$

# Restrict Operation

- Let  $\mathbf{x}$  be the root of  $\mathbf{G}_f$
- To compute  $\mathbf{G}_{f|_{y=b}}$  we consider:

**CASE2:  $x > y$**

$$\blacksquare \mathbf{f}|_{y=b} = \mathbf{f}$$

# Restrict Operation

- Let  $\mathbf{x}$  be the root of  $\mathbf{G}_f$
- To compute  $\mathbf{G}_f|_{y=b}$  we consider:  
**CASE2:  $\mathbf{x} < \mathbf{y}$** 
  - $\mathbf{f}|_{y=b} = (\neg \mathbf{x} \wedge (\mathbf{f}|_{x=0})|_{y=b}) \vee (\mathbf{x} \wedge (\mathbf{f}|_{x=1})|_{y=b})$
- We have to solve now two **smaller** problems!



# Algorithm for Restrict

**Algorithm Restrict(u,i,b)**

**Function Res(u)**

**if var(u) > i then return u**

**else if var(u) < i then**

**return mk(var(u),Res(low(u)),Res(high(u)))**

**else /\* var(u) = i \*/**

**if b = 0 then**

**return Res(low(u))**

**else /\* var(u) = i and b = 1 \*/**

**return Res(high(u))**

**return Res(u)**

*running time* =  $O(2^n)$ . Why?

# Efficient algorithm for Restrict

**Algorithm Restrict(u,i,b)**

**init( $G_{res}$ )**

**Function Res(u)**

**if  $G_{res}(u) \neq \text{empty}$  then return  $G_{res}(u)$**

**if  $\text{var}(u) > i$  then return  $u$**

**else if  $\text{var}(u) < i$  then**

**$r = \text{mk}(\text{var}(u), \text{Res}(\text{low}(u)), \text{Res}(\text{high}(u)))$**

**else /\*  $\text{var}(u) = \text{var}(v)$  \*/**

**if  $b = 0$  then**

**$r = \text{Res}(\text{low}(u))$**

**else /\*  $\text{var}(u) = \text{var}(v)$  and  $b = 1$  \*/**

**$r = \text{Res}(\text{high}(u))$**

**$G_{res}(u) = r$**

**return  $r$**

**return Res(u)**

*running time* =  $O(|G_u|)$ . Why?

# Quantification

- Extend the boolean language with

$$\exists \mathbf{x}.t \mid \forall \mathbf{x}.t$$

- They can be defined in terms of ROBDD operations:

$$\exists \mathbf{x}.t = t[0/\mathbf{x}] \vee t[1/\mathbf{x}]$$

$$\forall \mathbf{x}.t = t[0/\mathbf{x}] \wedge t[1/\mathbf{x}]$$

We can use an appropriate combination of *Restrict* and *Apply*

# Symbolic CTL Model Checking

- Represent the required **subsets of states** as boolean functions and hence as **ROBDDs**.
- Represent the **transition relation** as a boolean function and hence as a **ROBDD**.
- Reduce the iterative **fixed point computations** of the model checking process to **operations on OBDDs**.
- Check for the **termination** of the **fixpoint** computation by checking **ROBDD equivalence**.

# Symbolic Model Checking

- $\mathbf{K} = (\mathbf{S}, \mathbf{S}_0, \mathbf{R}, \mathbf{AP}, \mathbf{L})$
- Assume that if  $\mathbf{L}(s) = \mathbf{L}(s')$  then  $s = s'$ .
  - If not, *add* a few *new atomic propositions* if necessary, so as to distinguish states only based on the labeling.
- $\mathbf{AP} = \{p, q, r\}$
- $\mathbf{L}(s) = \{p\}$ 
  - $\mathbf{f}_s = p \wedge \neg q \wedge \neg r$
- $\mathbf{f}_{\{s_1, s_2, s_5\}} = \mathbf{f}_{s_1} \vee \mathbf{f}_{s_2} \vee \mathbf{f}_{s_5}$

# Symbolic Model Checking

- $\mathbf{K} = (\mathbf{S}, \mathbf{S}_0, \mathbf{R}, \mathbf{AP}, \mathbf{L})$
- $\mathbf{AP} = \{\mathbf{p}, \mathbf{q}, \mathbf{r}\}$
- *Add* the next-state boolean variables  $\{\mathbf{p}', \mathbf{q}', \mathbf{r}'\}$
- Suppose  $(s_1, s_2)$  in  $\mathbf{R}$  (i.e.  $\mathbf{R}(s_1, s_2)$ )  
with  $\mathbf{L}(s_1) = \{\mathbf{p}, \mathbf{q}\}$  and  $\mathbf{L}(s_2) = \{\mathbf{r}\}$ .

Then  $\mathbf{f}_{\mathbf{R}(s_1, s_2)} = \mathbf{f}_{s_1} \wedge \mathbf{f}'_{s_2}$ .

– where  $\mathbf{f}_{s_1} = \mathbf{p} \wedge \mathbf{q} \wedge \neg \mathbf{r}$  and  $\mathbf{f}'_{s_2} = \neg \mathbf{p}' \wedge \neg \mathbf{q}' \wedge \mathbf{r}'$

- $\mathbf{f}_{\mathbf{R}} = \bigvee_{(s_1, s_2) \in \mathbf{R}} (\mathbf{f}_{\mathbf{R}(s_1, s_2)})$
- Choose the ordering  $\mathbf{p} < \mathbf{p}' < \mathbf{q} < \mathbf{q}' < \mathbf{r} < \mathbf{r}'$  !

# CTL symbolic Model Checking

- $||[\mathbf{x}_i]|| = \mathbf{f}_{\mathbf{x}_i}(\mathbf{x}_1, \dots, \mathbf{x}_n) = \mathbf{x}_i$   
(the OBDD for the *boolean variable*  $\mathbf{x}_i$ )
- $||[\neg\phi]|| = \neg\mathbf{f}_{\phi}(\mathbf{x}_1, \dots, \mathbf{x}_n)$   
(apply negation to the OBDD for  $\phi$ )
- $||[\phi \vee \psi]|| = \mathbf{f}_{\phi}(\mathbf{x}_1, \dots, \mathbf{x}_n) \vee \mathbf{f}_{\psi}(\mathbf{x}_1, \dots, \mathbf{x}_n)$   
(apply  $\vee$  operation to the OBDDs for  $\phi$  and  $\psi$ )
- $||[\phi \wedge \psi]|| = \mathbf{f}_{\phi}(\mathbf{x}_1, \dots, \mathbf{x}_n) \wedge \mathbf{f}_{\psi}(\mathbf{x}_1, \dots, \mathbf{x}_n)$   
(apply  $\wedge$  operation to the OBDDs for  $\phi$  and  $\psi$ )

# CTL Symbolic Model Checking

- $\llbracket \mathbf{EX} \phi \rrbracket =$   
 $\exists \mathbf{x}'_1, \dots, \mathbf{x}'_n (\mathbf{f}_\phi(\mathbf{x}'_1, \dots, \mathbf{x}'_n) \wedge$   
 $\mathbf{f}_R(\mathbf{x}_1, \dots, \mathbf{x}_n, \mathbf{x}'_1, \dots, \mathbf{x}'_n))$

This is also called the *relational product*, or the *pre-image of  $\llbracket \phi \rrbracket$  by  $R$*  (see *Section 6.6* in *Clarke's book* for a more *efficient algorithm*).

- $\llbracket \mathbf{EU}(\phi, \psi) \rrbracket = \mu \mathbf{Z}. (\mathbf{f}_\psi(\mathbf{x}_1, \dots, \mathbf{x}_n) \vee$   
 $(\mathbf{f}_\phi(\mathbf{x}_1, \dots, \mathbf{x}_n) \wedge \mathbf{EX} \mathbf{Z}))$
- $\llbracket \mathbf{EG} \phi \rrbracket = \nu \mathbf{Z}. (\mathbf{f}_\phi(\mathbf{x}_1, \dots, \mathbf{x}_n) \wedge \mathbf{EX} \mathbf{Z})$



# Symbolic model checking: example

Let  $V = \{x_1, \dots, x_n\}$ , then  $\llbracket \mathbf{EG} \psi \rrbracket$  can be computed as follows:

1. Assume the ROBDD  $f_\psi(x_1, \dots, x_n)$  has been computed.
2. Set  $X_0 = f_\psi(x'_1, \dots, x'_n)$  [ computed from  $f_\psi(x_1, \dots, x_n)$  by *variable substitution* ]
3. We need to compute  $X_{i+1} = X_i \cap Y_i$  where:  
$$Y_i = \exists x'_1, \dots, x'_n (f_\psi(x'_1, \dots, x'_n) \wedge f_R(x_1, \dots, x_n, x'_1, \dots, x'_n))$$
  
 $X_{i+1}$  can easily be computed as  $X_i \wedge Y_i$
4. Check **whether**  $X_{i+1} = X_i$  by checking whether the corresponding ROBDDs are **identical**.
5. **If not**, substitute the *next-state* variables for the *state-variables* in  $X_{i+1}$ , and repeat from **step 3**.

## Algorithm Compute\_EG( $\beta$ )

$f_1(x) := f_\beta(x);$

$j=1;$

repeat

$j := j+1;$

$f_j := f_\beta(x) \wedge \exists x'.(f_R(x, x') \wedge f_{j-1}(x'));$

until  $f_j(x) = f_{j-1}(x);$

## Algorithm Compute\_EU( $\beta_1, \beta_2$ )

$f_1(x) := f_{\beta_2}(x);$

$j=1;$

repeat

$j := j+1;$

$f_j := f_{\beta_2}(x) \vee (f_{\beta_1}(x) \wedge \exists x'.(f_R(x, x') \wedge f_{j-1}(x')));$

until  $f_j(x) = f_{j-1}(x);$

# CTL Symbolic model checking

Finally, assuming boolean variables  $V = \{x_1, \dots, x_n\}$ , and the ROBDD for  $[[\phi]]$  already computed.

- Checking whether

$$\mathbf{K} \models \phi$$

amounts to checking whether the ROBDD for  $\mathbf{f}_{\text{Init}} \wedge \mathbf{f}_{\neg\phi}$  is **identical** to the ROBDD for  $\mathbf{0}$ , where  $\mathbf{f}_{\text{Init}}$  is the ROBDD for the set  $[[\text{Init}]]$  of **initial states** of  $\mathbf{K}$ .

(recall that  $\mathbf{K} \models \phi$  iff  $[[\text{Init}]] \subseteq [[\phi]]$  iff  $[[\text{Init}]] \cap [[\neg\phi]] = \emptyset$ )

# Pre-image computation with BDD

Let us consider the *Pre-image* operation

$$\exists x'_1, \dots, x'_n (f_\psi(x'_1, \dots, x'_n) \wedge f_R(x_1, \dots, x_n, x'_1, \dots, x'_n))$$

*Pre-image* is a special case of the *Relational Product*

$$\exists x' (R_1(x, x') \wedge R_2(x', z))$$

where  $R_1$  is  $R$ ,  $R_2$  is  $\psi$  and  $z$  is empty.

*Pre-image* can easily be computed by applying  $\wedge$  to the BDD's for  $\psi$  and  $R$ , and then existential elimination of the primed variables.

However, the intermediate BDD for

$$f_\psi(x'_1, \dots, x'_n) \wedge f_R(x_1, \dots, x_n, x'_1, \dots, x'_n)$$

is usually far bigger than the final result.

This can be avoided by exploiting *early quantification*, whenever possible.

# Pre-image computation with BDD

*Early quantification* is based on the fact that:

- If  $\mathbf{x}_1 < \mathbf{x}_2$  and the top variable of  $\mathbf{f}$  is  $\mathbf{x}_1$  then

$$\exists \mathbf{x}_2 (\mathbf{x}_1 \rightarrow \mathbf{f}|_{\mathbf{x}_1=1}, \mathbf{f}|_{\mathbf{x}_1=0}) \equiv (\mathbf{x}_1 \rightarrow \exists \mathbf{x}_2 \mathbf{f}|_{\mathbf{x}_1=1}, \exists \mathbf{x}_2 \mathbf{f}|_{\mathbf{x}_1=0})$$

(recall that  $\exists \mathbf{x} (\mathbf{f} \text{ op } \mathbf{g}) \equiv \mathbf{f} \text{ op } \exists \mathbf{x} \mathbf{g}$ , whenever  $\mathbf{f}$  does not depend on  $\mathbf{x}$ )

- If the top variable of  $\mathbf{g}$  is  $\mathbf{x}_2$  then

$$\exists \mathbf{x}_2 (\mathbf{x}_2 \rightarrow \mathbf{f}|_{\mathbf{x}_2=1}, \mathbf{f}|_{\mathbf{x}_2=0}) \equiv (\mathbf{f}|_{\mathbf{x}_2=1} \vee \mathbf{f}|_{\mathbf{x}_2=0})$$

This means that we can devise an algorithm that computes the *pre-image* by applying quantification as soon as it is possible.

This avoids computing the *conjunction* (which is usually bigger than the final result) during the computation of the *pre-image*.

# Pre-image computation with BDD

**Algorithm** RelationalProduct( $u, v, \mathcal{I}$ ) /\*  $\exists \mathcal{I} (f^u \wedge f^v)$  \*/

**init**( $G$ )

**Function** RelPrd( $u, v, \mathcal{I}$ )

**if**  $u = 0$  or  $v = 0$  **then return** 0

**if**  $u = 1$  and  $v = 1$  **then return** 1

**if**  $G(u, v) \neq \text{NIL}$  **then return**  $G(u, v)$

$z = \min(\text{var}(u), \text{var}(v))$

**if**  $\text{var}(v) = \text{var}(u)$  **then**

$r_1 = \text{RelPrd}(\text{high}(u), \text{high}(v), \mathcal{I})$  ;  $r_2 = \text{RelPrd}(\text{low}(u), \text{low}(v), \mathcal{I})$

**else if**  $z = \text{var}(u)$  **then**

$r_1 = \text{RelPrd}(\text{high}(u), v, \mathcal{I})$  ;  $r_2 = \text{RelPrd}(\text{low}(u), v, \mathcal{I})$

**else** /\*  $z = \text{var}(v)$  \*/

$r_1 = \text{RelPrd}(u, \text{high}(v), \mathcal{I})$  ;  $r_2 = \text{RelPrd}(u, \text{low}(v), \mathcal{I})$

**if**  $z \notin \mathcal{I}$  **then** /\*  $z$  is not a quantified variable \*/

$r = \text{mk}(z, r_1, r_2)$

**else** /\*  $z \in \mathcal{I}$  :  $z$  is a quantified variable \*/

$r = \text{Apply}(v, r_1, r_2)$

$G(u, v) = r$

**return**  $r$

**return** RelPrd( $u, v, \mathcal{I}$ )

# Symbolic Model Checking

- The actual Kripke structure will be, in general, too large.
  - **State explosion.**
- So one must try to **compute the ROBDDs directly from the system model (NuSMV program)** and run the model checking procedure with the help of this **implicit** representation.
  - **Symbolic model checking.**
- This may not be sufficient, though! **Additional techniques may be needed** (e.g., **abstraction**).