

# Binary Decision Diagrams

(Based on [Clarke et al. 1999] and [Bryant 1986])

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# Boolean Functions

🌐 Boolean functions are widely used in

- ☀️ digital logic design and testing,
- ☀️ artificial intelligence,
- ☀️ combinatorics, and
- ☀️ model checking.

🌐 Boolean operators

- ☀️ Conjunction (and):  $x \cdot y$  ( $x \wedge y$ )
- ☀️ Disjunction (or):  $x + y$  ( $x \vee y$ )
- ☀️ Negation (not):  $\bar{x}$  ( $\neg x$ )
- ☀️ Equivalence (if and only if):  $\leftrightarrow$

🌐 Example:  $f(x_1, x_2, x_3, x_4) = (x_1 \leftrightarrow x_2) \cdot (x_3 \leftrightarrow x_4)$

# Representations of Boolean Functions

- 🌐 A variety of methods had earlier been developed for representing and manipulating Boolean functions:
  - ☀ Truth table
  - ☀ Karnaugh map
  - ☀ Sum-of-products form
  - ☀ Binary decision tree
- 🌐 These representations are quite impractical, because every function of  $n$  arguments has a representation of size  $2^n$  or more.

# Truth Table

A truth table for  $f(x_1, x_2, x_3, x_4) = (x_1 \leftrightarrow x_2) \cdot (x_3 \leftrightarrow x_4)$ .

$x_1$	$x_2$	$x_3$	$x_4$	$f$
0	0	0	0	1
0	0	0	1	0
0	0	1	0	0
0	0	1	1	1
0	1	0	0	0
0	1	0	1	0
0	1	1	0	0
0	1	1	1	0

$x_1$	$x_2$	$x_3$	$x_4$	$f$
1	0	0	0	0
1	0	0	1	0
1	0	1	0	0
1	0	1	1	0
1	1	0	0	1
1	1	0	1	0
1	1	1	0	0
1	1	1	1	1

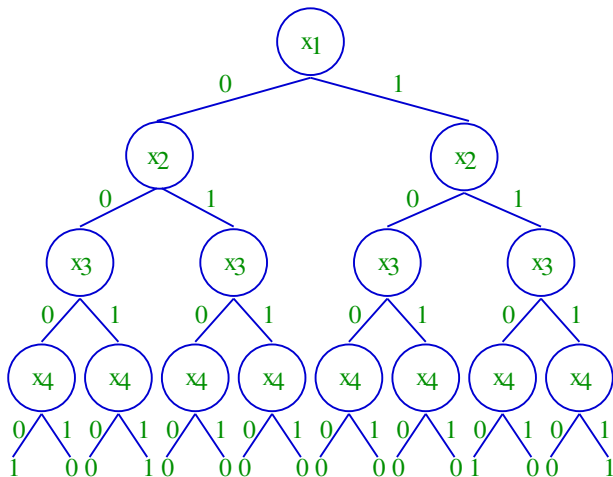
# Karnaugh Map

A Karnaugh table for  $f(x_1, x_2, x_3, x_4) = (x_1 \leftrightarrow x_2) \cdot (x_3 \leftrightarrow x_4)$ .

$x_3x_4$ \ $x_1x_2$	00	01	11	10
00	1	0	1	0
01	0	0	0	0
11	1	0	1	0
10	0	0	0	0

# Binary Decision Tree

A binary decision tree for  $f(x_1, x_2, x_3, x_4) = (x_1 \leftrightarrow x_2) \cdot (x_3 \leftrightarrow x_4)$ .



- 🌐 More practical approaches utilize representations that, at least for many functions, are not of exponential size.
  - ☀ reduced sum of products
  - ☀ factored into unate (cf. monotone) functions
- 🌐 These representations still suffer from several drawbacks:
  - ☀ Certain common functions require representations of exponential size.
  - ☀ Performing a simple operation could yield a function with an exponential representation.
  - ☀ None of these representations are *canonical forms* (which are convenient for equivalence testing).

# Binary Decision Diagrams

- 🌐 A **binary decision diagram (BDD)** represents a Boolean function as a rooted, directed acyclic graph (function graph).
- 🌐 We use  $r(G)$  to denote the root of a function graph  $G$ .
- 🌐 The vertex set  $V$  of a function graph  $G$  contains two types of vertices.
  - ☀️ A **nonterminal** vertex  $v$  has
    - 👤 an argument index  $index(v) \in \{1, \dots, n\}$  and
    - 👤 two children  $low(v), high(v) \in V$ .
  - ☀️ A **terminal** vertex  $v$  has a value  $value(v) \in \{0, 1\}$ .

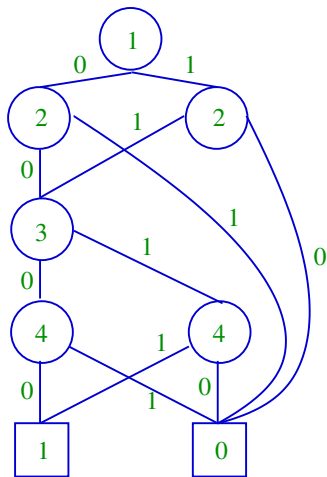
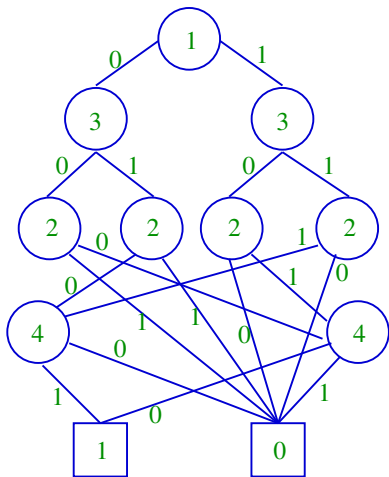


# Ordered Binary Decision Diagrams

- 🌐 An **ordered binary decision diagram (OBDD)** is defined by imposing a total ordering over the nonterminal vertices.
  - ☀️ For any nonterminal vertex  $v$ ,
    - 👤 if  $low(v)$  is nonterminal, then we must have  $index(v) < index(low(v))$ ;
    - 👤 if  $high(v)$  is nonterminal, then we must have  $index(v) < index(high(v))$ .
- 🌐 Further minimality conditions will be introduced later.
- 🌐 OBDDs are representations of Boolean functions with *canonical forms* and *reasonable size*.
- 🌐 The size of the graph is highly sensitive to arguments ordering.

# Ordering

Two OBDDs for  $f(x_1, x_2, x_3, x_4) = (x_1 \leftrightarrow x_2) \cdot (x_3 \leftrightarrow x_4)$  with different orderings.



# Notations

- All functions have the same  $n$  arguments:  $x_1, \dots, x_n$ .
- A **restriction** of  $f$  is denoted  $f|_{x_i=b}$  where  $b$  is a constant.

$$f|_{x_i=b}(x_1, \dots, x_n) = f(x_1, \dots, x_{i-1}, b, x_{i+1}, \dots, x_n)$$

- A **composition** of  $f$  and  $g$  is denoted  $f|_{x_i=g}$  where  $g$  is a Boolean function.

$$f|_{x_i=g}(x_1, \dots, x_n) = f(x_1, \dots, x_{i-1}, g(x_1, \dots, x_n), x_{i+1}, \dots, x_n)$$

## Notations (cont.)

- The **Shannon expansion** of a function around variable  $x_i$  is given by:

$$f = x_i \cdot f|_{x_i=1} + \bar{x}_i \cdot f|_{x_i=0}$$

- The **dependency set** of a function  $f$  is denoted  $I_f$ .

$$I_f = \{i \mid f|_{x_i=0} \neq f|_{x_i=1}\}$$

- The **satisfying set** of a function  $f$  is denoted  $S_f$ .

$$S_f = \{(x_1, \dots, x_n) \mid f(x_1, \dots, x_n) = 1\}$$

# Correspondence

- 🌐 A function graph (OBDD)  $G$  having root vertex  $v$  denotes a function  $f_v$  defined recursively as follows:
  - ☀️ If  $v$  is a terminal vertex:
    - 👁️ If  $value(v) = 1$ , then  $f_v = 1$ .
    - 👁️ If  $value(v) = 0$ , then  $f_v = 0$ .
  - ☀️ If  $v$  is a nonterminal vertex with  $index(v) = i$ , then  $f_v$  is the function

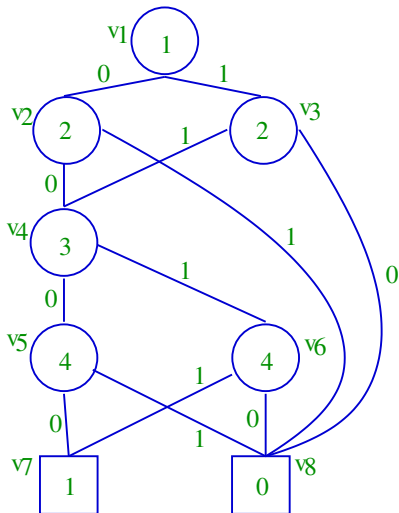
$$f_v(x_1, \dots, x_n) = \bar{x}_i \cdot f_{low(v)}(x_1, \dots, x_n) + x_i \cdot f_{high(v)}(x_1, \dots, x_n).$$

# Correspondence (cont.)

- 🌐 A path in the graph starting from the root is defined by a set of argument values.
- 🌐 The value of the function for these arguments equals the value of the terminal vertex at the end of the path.
- 🌐 Every vertex in the graph is contained in at least one path.

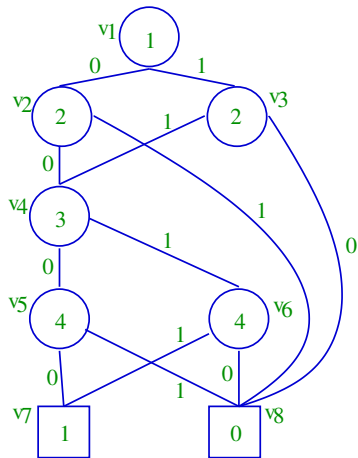
## Correspondence (cont.)

$$\begin{aligned}
 f_{v_8} &= 0 \\
 f_{v_7} &= 1 \\
 f_{v_6} &= \bar{x}_4 \cdot f_{v_8} + x_4 \cdot f_{v_7} \\
 &= x_4 \\
 f_{v_5} &= \bar{x}_4 \cdot f_{v_7} + x_4 \cdot f_{v_8} \\
 &= \bar{x}_4 \\
 f_{v_4} &= \bar{x}_3 \cdot f_{v_5} + x_3 \cdot f_{v_6} \\
 &= \bar{x}_3 \cdot \bar{x}_4 + x_3 \cdot x_4 \\
 &\dots \\
 &\dots \\
 &\dots \\
 f_{v_1} &= (\bar{x}_1 \cdot \bar{x}_2 + x_1 \cdot x_2) \cdot (\bar{x}_3 \cdot \bar{x}_4 + x_3 \cdot x_4)
 \end{aligned}$$



# Subgraph

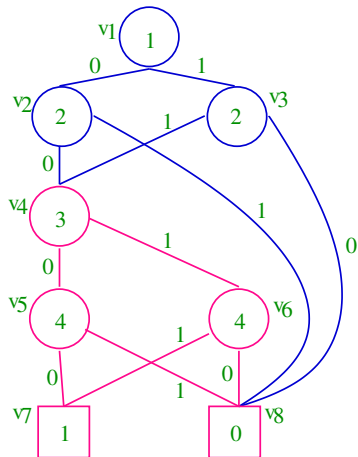
- For any vertex  $v$  in a function graph  $G$ , the **subgraph** rooted at  $v$ , denoted by  $sub(G, v)$  is defined as the graph consisting of  $v$  and all its descendants.





# Subgraph

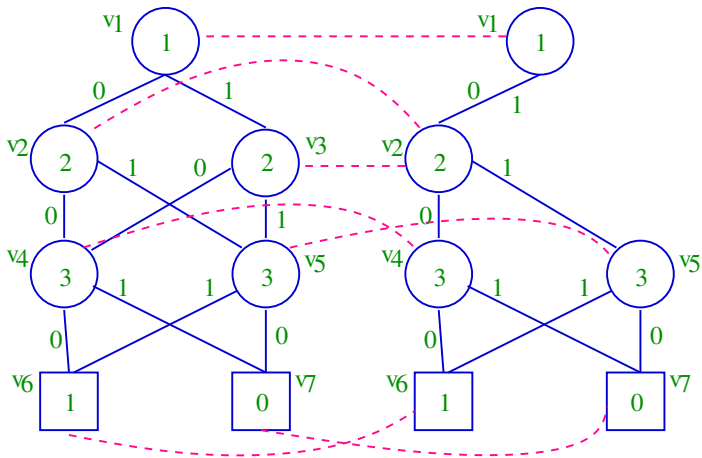
- For any vertex  $v$  in a function graph  $G$ , the **subgraph** rooted at  $v$ , denoted by  $sub(G, v)$  is defined as the graph consisting of  $v$  and all its descendants.



# Isomorphism

- 🌐 Function graphs  $G$  and  $G'$  are **isomorphic**, denoted by  $G \sim G'$ , if there exists a **one-to-one** function  $\sigma$  from vertices of  $G$  onto the vertices of  $G'$  such that for any vertex  $v$  if  $\sigma(v) = v'$ , then either
- ☀ both  $v$  and  $v'$  are terminal vertices with  $value(v) = value(v')$ ,  
or
  - ☀ both  $v$  and  $v'$  are nonterminal vertices with  
 $index(v) = index(v')$ ,  $\sigma(low(v)) = low(v')$ , and  
 $\sigma(high(v)) = high(v')$

# Isomorphism (cont.)



Is this an isomorphic mapping? (part of it is)

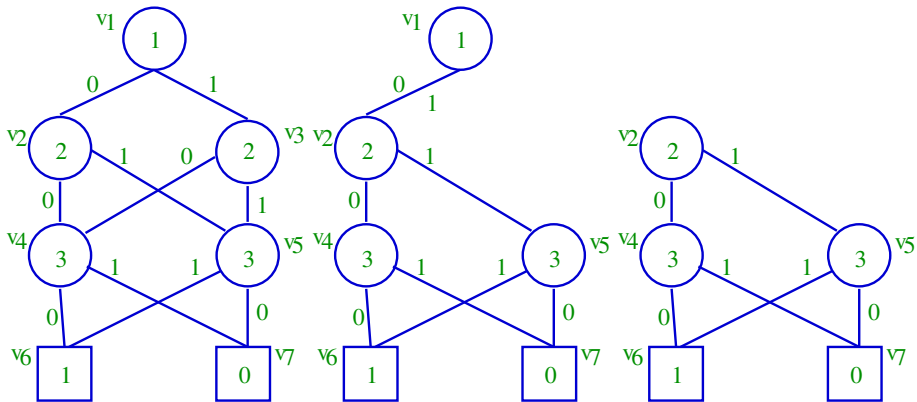
# Isomorphism (cont.)


- 🌐 The isomorphic mapping  $\sigma$  is quite constrained:
  - ☀️  $r(G)$  must map to the  $r(G')$ ,
  - ☀️  $low(r(G))$  must map to  $low(r(G'))$ ,
  - ☀️ and so on all the way down to the terminal vertices.
- 🌐 Lemma 1: If  $G$  is isomorphic to  $G'$  by mapping  $\sigma$ , denoted by  $G \sim_{\sigma} G'$ , then for any vertex  $v$  in  $G$ ,  $sub(G, v) \sim sub(G', \sigma(v))$ .

# Reduced Function Graph

- 🌐 A function graph  $G$  is **reduced** if
  - ☀️ it contains no vertex  $v$  with  $low(v) = high(v)$ ,
  - ☀️ nor does it contain distinct vertices  $v$  and  $v'$  such that the subgraphs rooted by  $v$  and  $v'$  are isomorphic.
- 🌐 A reduced function graph is now commonly called (Reduced) OBDD.
- 🌐 Lemma 2: For every vertex  $v$  in a reduced function graph  $G$ ,  $sub(G, v)$  is itself a reduced function graph.

# Reduced Function Graph (cont.)



-  Theorem: For any Boolean function  $f$ , there is a unique (up to isomorphism) reduced function graph denoting  $f$  and any other function graph denoting  $f$  contains more vertices.

# Basic Operations

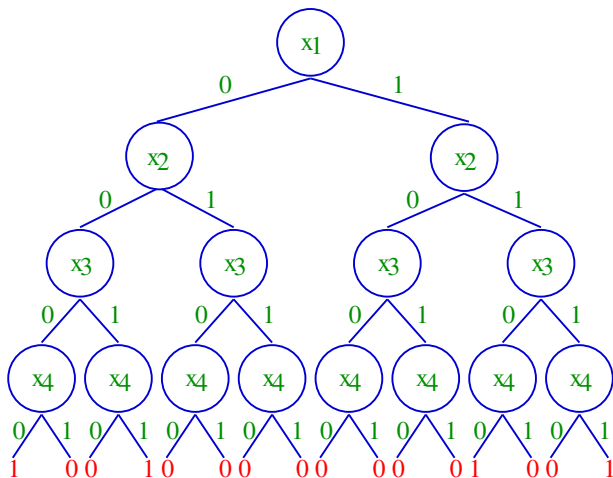
Procedure	Result	Time Complexity
Reduce	$G$ reduced to canonical form	$O( G  \cdot \log  G )$
Apply	$f_1 \langle op \rangle f_2$	$O( G_1  \cdot  G_2 )$
Restrict	$f \mid_{x_i=b}$	$O( G  \cdot \log  G )$
Compose	$f_1 \mid_{x_i=f_2}$	$O( G_1 ^2 \cdot  G_2 )$
Satisfy-one	some element of $S_f$	$O(n)$
Satisfy-all	$S_f$	$O(n \cdot  S_f )$
Satisfy-count	$ S_f $	$O( G )$



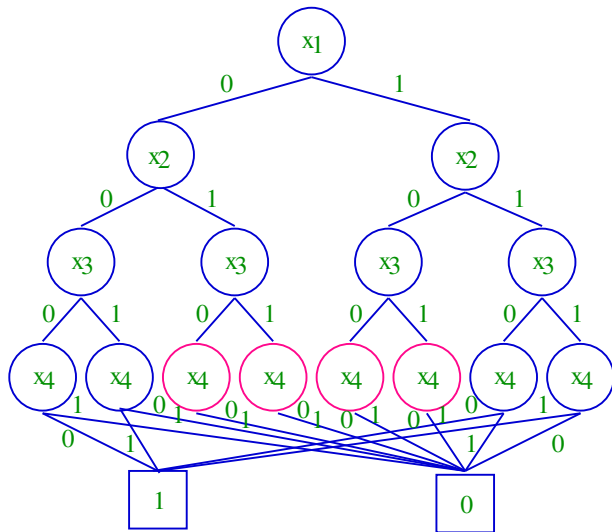
# Reduction

- 🌐 The *reduction* algorithm transforms an arbitrary function graph into a reduced graph denoting the same function.
- 🌐 The algorithm works from the terminal vertices up to the root:
  - ☀️ Remove duplicate terminals (terminal vertices  $v$  and  $u$  such that  $value(v) = value(u)$ ).
  - ☀️ Remove duplicate nonterminals (nonterminal vertices  $v$  and  $u$  such that  $index(v) = index(u)$ ,  $id(low(v)) = id(low(u))$ , and  $id(high(v)) = id(high(u))$ ).
  - ☀️ Remove duplicate tests (a nonterminal vertex  $v$  such that  $low(v) = high(v)$ ).

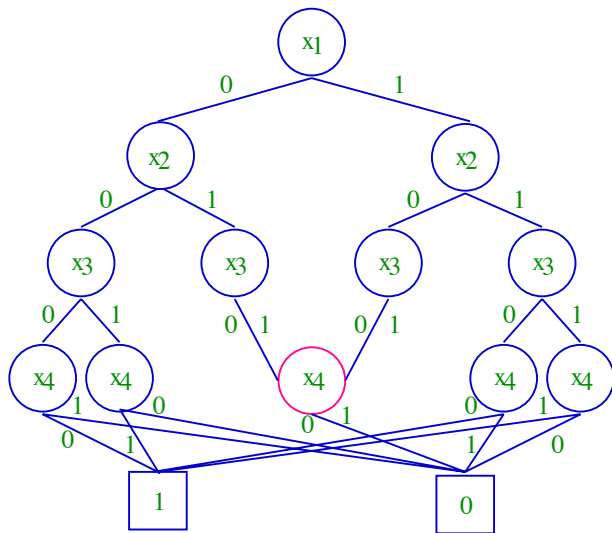
# A Reduction Example



# A Reduction Example

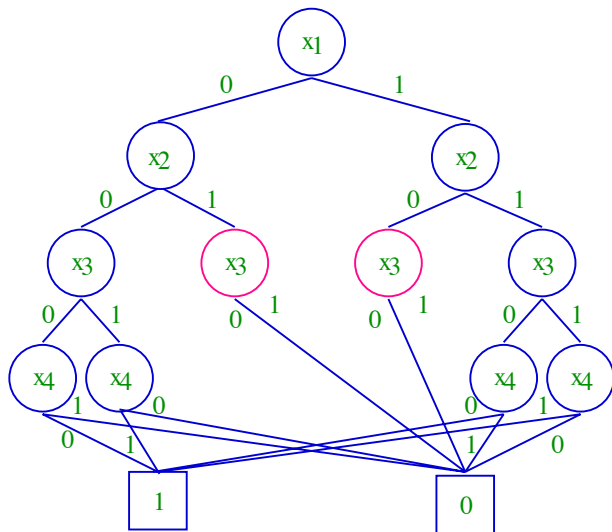


# A Reduction Example



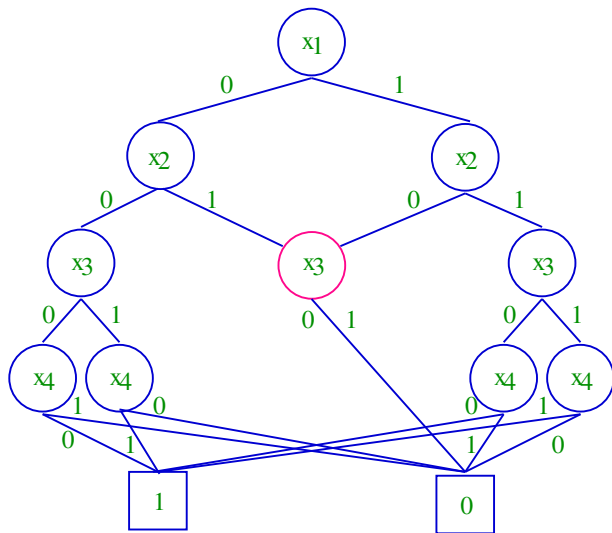
Note: not strictly bottom to top (for better layouts).

# A Reduction Example



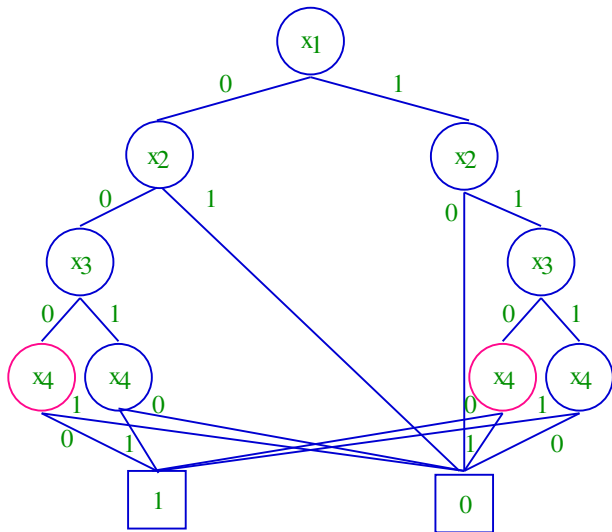
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# A Reduction Example

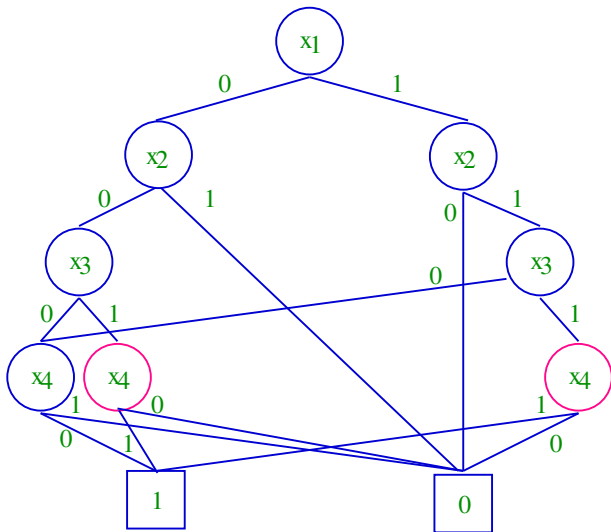


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# A Reduction Example

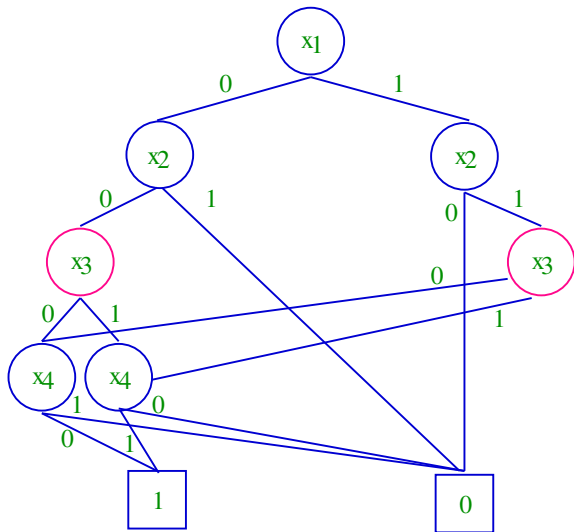


# A Reduction Example

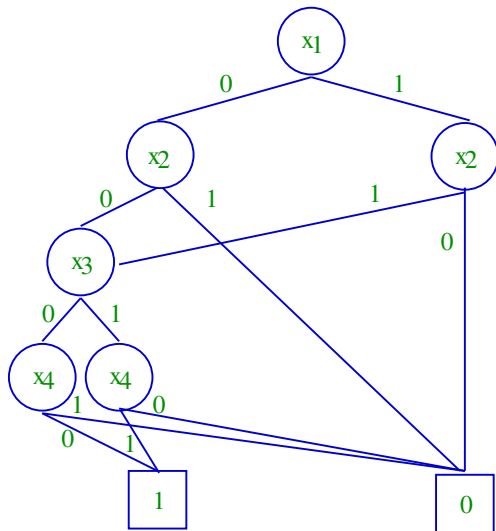




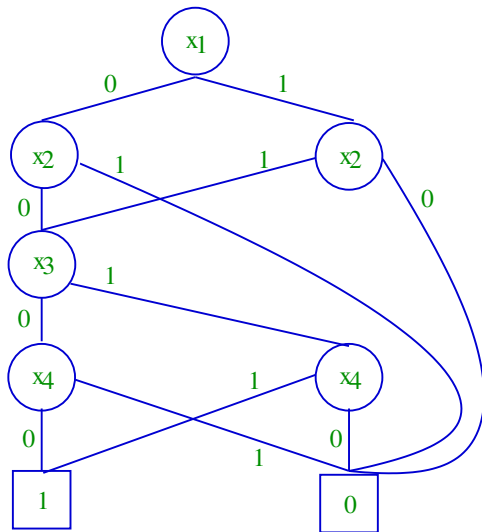
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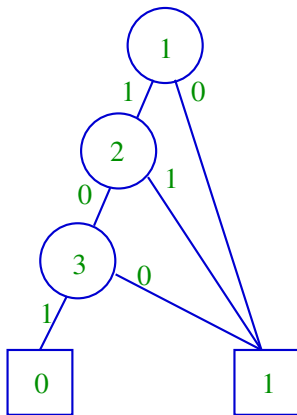


# Restriction

- 🌐 The procedure *Restrict* transforms the graph representing a function  $f$  into one representing the function  $f|_{x_i=b}$ .
- 🌐 Steps of *Restrict*:
  - ☀️ Look for a vertex  $v$  with  $index(v) = i$ .
  - ☀️ Change it to point either to  $low(v)$  (for  $b = 0$ ) or to  $high(v)$  (for  $b = 1$ ).
  - ☀️ After changing every vertex  $v$  with  $index(v) = i$ , run the reduction procedure.

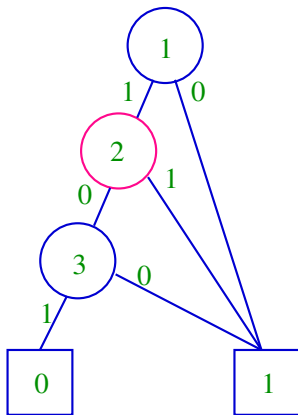
# A Restriction Example

$$\overline{x_1 \cdot \overline{x_2} \cdot x_3} \Big|_{x_2=0} = \overline{x_1 \cdot x_3}$$



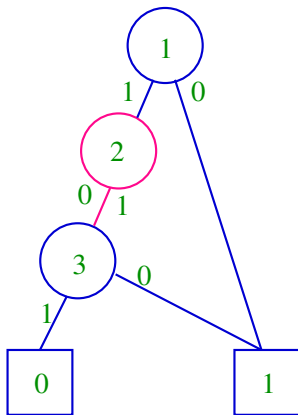
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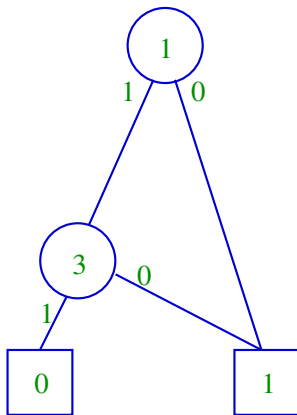
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$$\overline{x_1 \cdot \bar{x}_2 \cdot x_3} \Big|_{x_2=0} = \overline{x_1 \cdot x_3}$$





- 🌐 The procedure *Apply* takes graphs representing functions  $f_1$  and  $f_2$ , a binary operator  $\langle op \rangle$ , and produces a reduced graph representing the function  $f_1 \langle op \rangle f_2$  defined as:

$$[f_1 \langle op \rangle f_2](x_1, \dots, x_n) = f_1(x_1, \dots, x_n) \langle op \rangle f_2(x_1, \dots, x_n).$$

- 🌐 It is based on the following recursion derived from the Shannon expansion:

$$f_1 \langle op \rangle f_2 = \bar{x}_i \cdot (f_1 |_{x_i=0} \langle op \rangle f_2 |_{x_i=0}) + x_i \cdot (f_1 |_{x_i=1} \langle op \rangle f_2 |_{x_i=1})$$

# Apply (cont.)

🌐 Given function  $f_1$  rooted at  $v_1$  and function  $f_2$  rooted at  $v_2$ , there are four cases to consider:

- ☀️  $v_1$  and  $v_2$  are terminals:  $f_1 \langle op \rangle f_2 = value(v_1) \langle op \rangle value(v_2)$
- ☀️  $index(v_1) = index(v_2)$ : use the derived recursion
- ☀️  $index(v_1)(= i) < index(v_2)$ :  $f_2|_{x_i=0} = f_2|_{x_i=1} = f_2$ , so

$$f_1 \langle op \rangle f_2 = \bar{x}_i \cdot (f_1|_{x_i=0} \langle op \rangle f_2) + x_i \cdot (f_1|_{x_i=1} \langle op \rangle f_2)$$

- ☀️  $index(v_1) > index(v_2)$ : analogously as above

🌐 To avoid repeating the operation on two same nodes, we use dynamic programming.

## Apply (cont.)

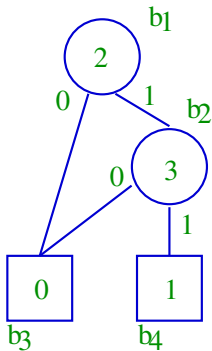
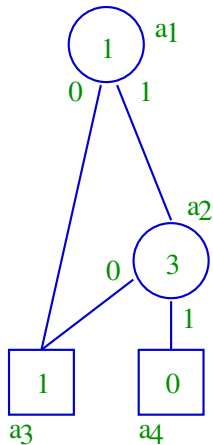
```
function Apply(v1, v2: vertex  $\langle op \rangle$ : operator): vertex  
{var T: array[1..|G1|, 1..|G2|] of vertex;}  
begin  
    Initialize all elements of T to null;  
    u := Apply-step(v1, v2);  
    return(Reduce(u));  
end;
```

# Apply (cont.)

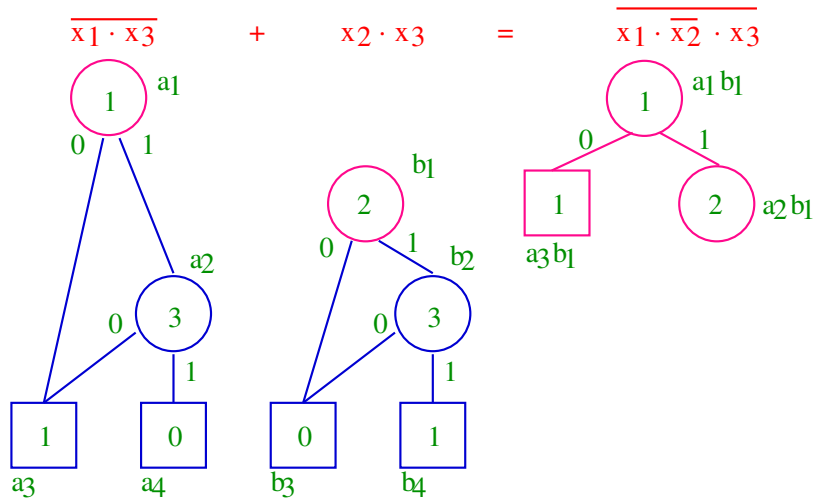
```
function Apply-step(v1, v2: vertex): vertex;
begin
  u := T[v1.id, v2.id];
  if u ≠ null then return(u); {have already evaluated}
  u := new vertex record; u.mark := false;
  T[v1.id, v2.id] := u; {add vertex to table}
  u.value := v1.value ⟨op⟩ v2.value;
  if u.value ≠ X
    then u.index := n + 1; u.low := null; u.high := null;
  else {create nonterminal and evaluate further down}
    u.index := Min(v1.index, v2.index);
    if v1.index = u.index
      then begin vlow1 := v1.low; vhigh1 := v1.high end
      else begin vlow1 := v1; vhigh1 := v1 end;
    if v2.index = u.index
      then begin vlow2 := v2.low; vhigh2 := v2.high end
      else begin vlow2 := v2; vhigh2 := v2 end;
    u.low := Apply-step(ulow1, vlow2);
    u.high := Apply-step(vhigh1, vhigh2);
  return(u);
end;
```

# An Apply Example

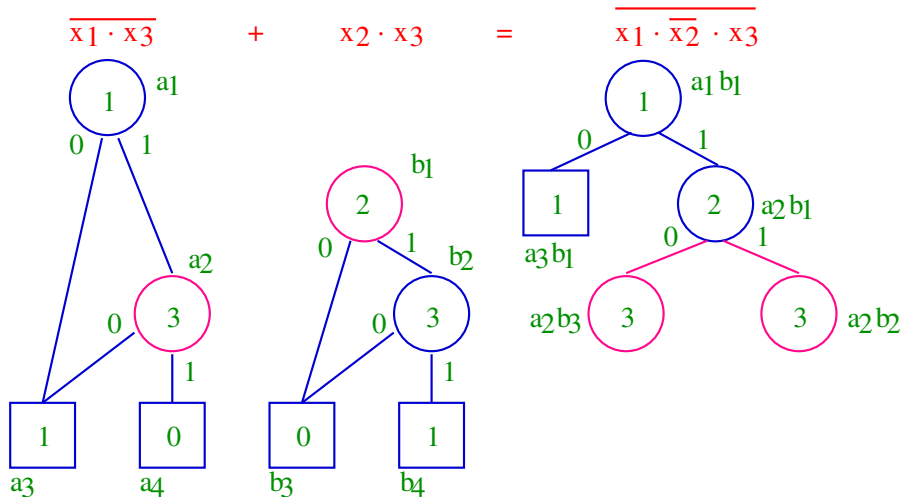
$$\overline{x_1 \cdot x_3} + x_2 \cdot x_3 = \overline{x_1 \cdot \overline{x_2} \cdot x_3}$$



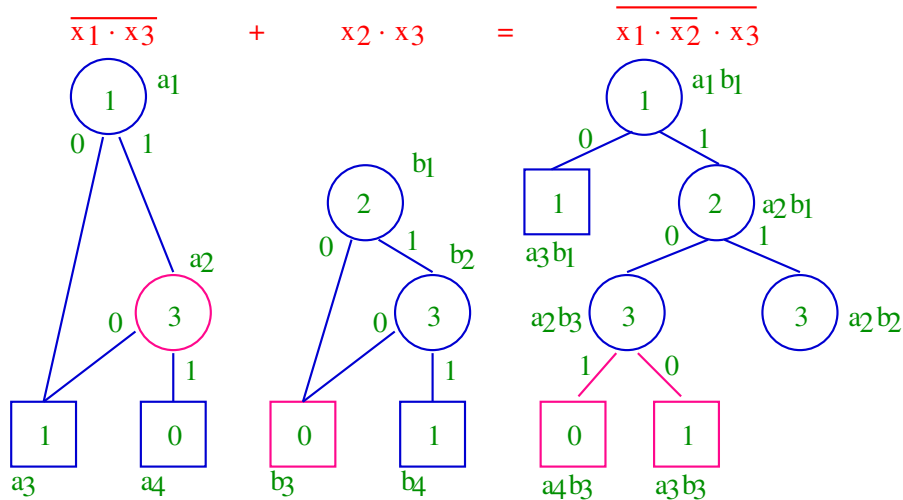
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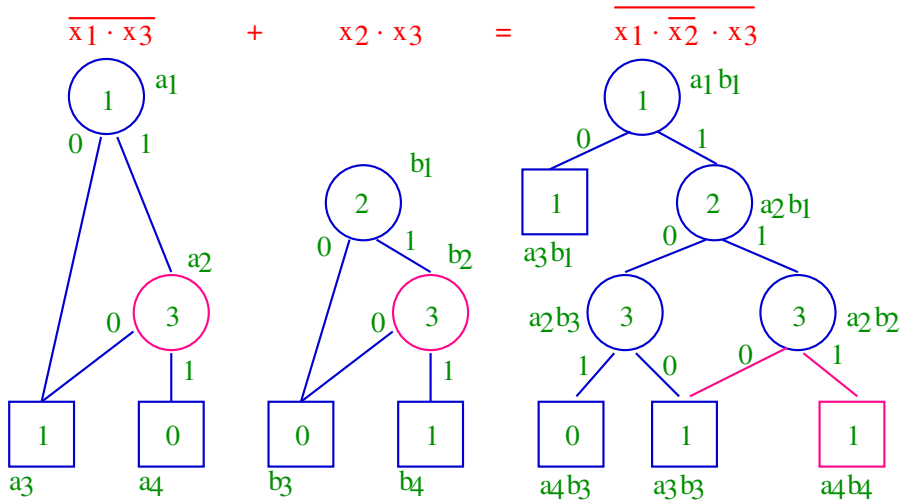


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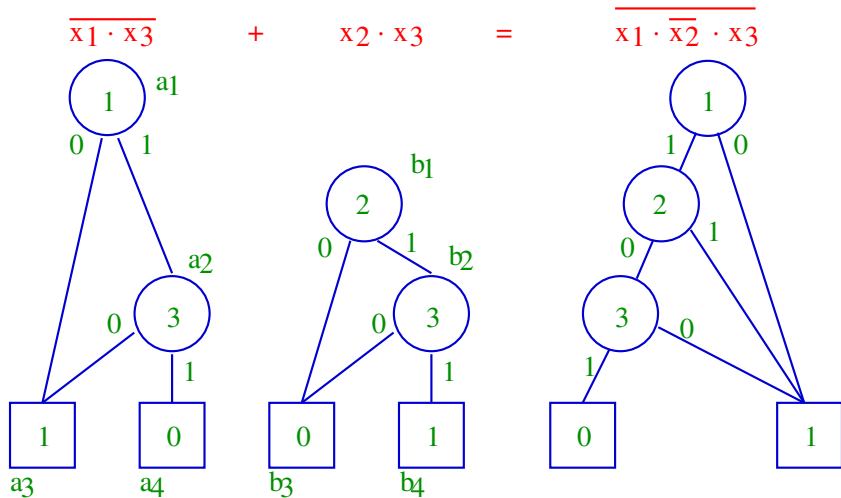




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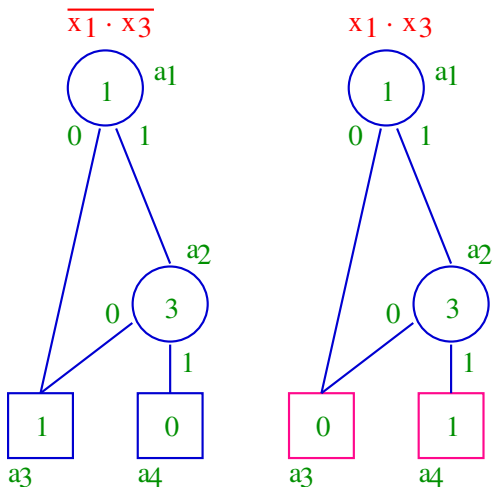


# An Apply Example



# Complementation

- 🌐 To complement an OBDD, simply complement its terminal vertices.



# Composition

- 🌐 The procedure *Compose* constructs the graph for the function obtained by composing two functions.
- 🌐 Composition can be expressed in terms of restriction and Boolean operations according to the following expansion:

$$f_1 |_{x_i=f_2} = f_2 \cdot f_1 |_{x_i=1} + (\neg f_2) \cdot f_1 |_{x_i=0}$$

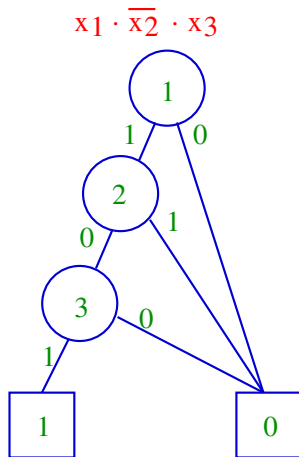
- 🌐 It is sufficient to use *Restrict* and *Apply* to implement *Compose*.

# Satisfy-one

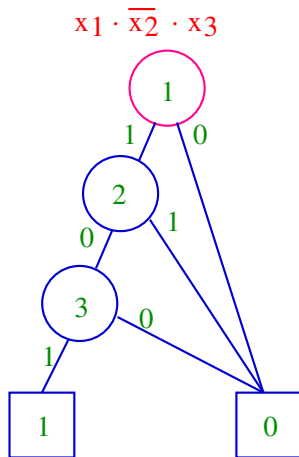
- The *Satisfy-one* procedure utilizes a classic depth-first search with backtracking.

```
function Satisfy-one(v: vertex; x: array[1..n] of integer): boolean
begin
  if value(v) = 0 then return false;
  if value(v) = 1 then return true;
  x[i] := 0;
  if Satisfy-one(low(v), x) then return true;
  x[i] := 1;
  return Satisfy-one(high(v), x);
end;
```

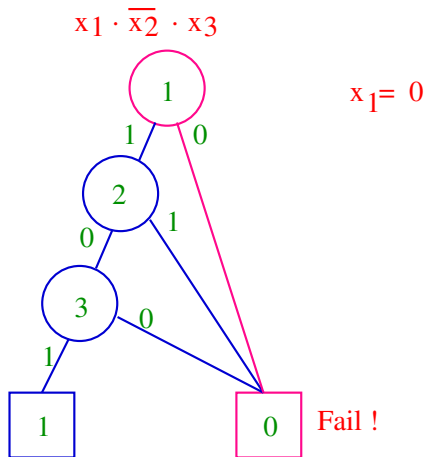
# A Satisfy-one Example



# A Satisfy-one Example

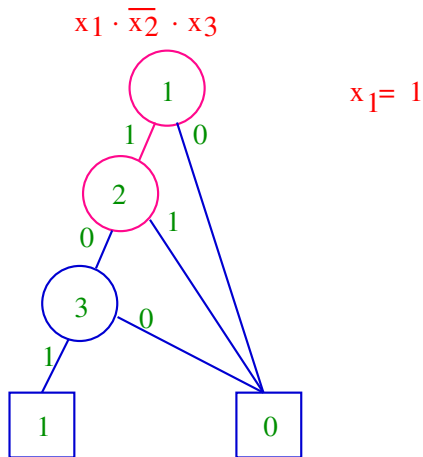


# A Satisfy-one Example

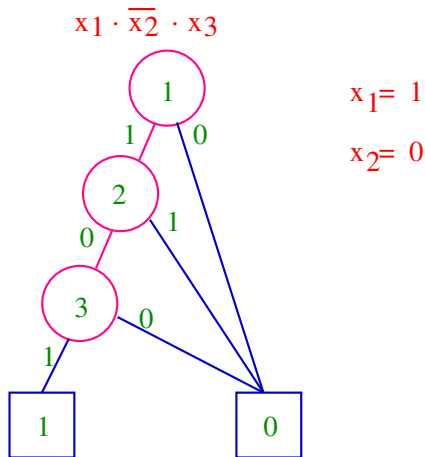




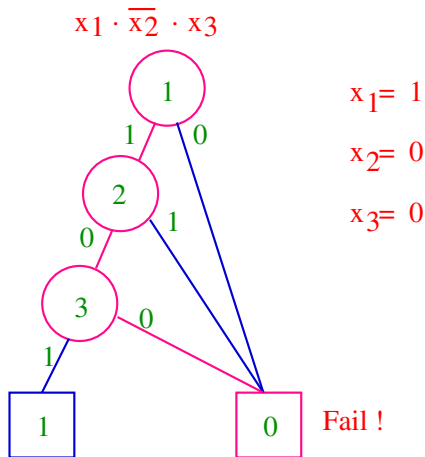
# A Satisfy-one Example



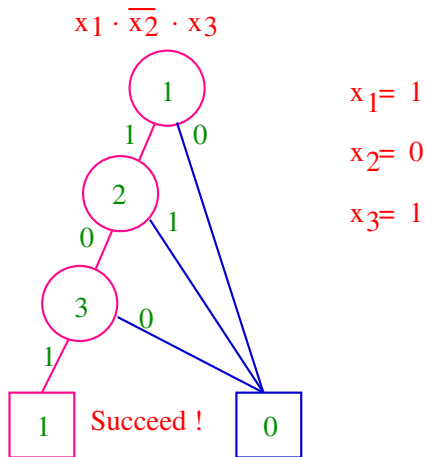
# A Satisfy-one Example



# A Satisfy-one Example

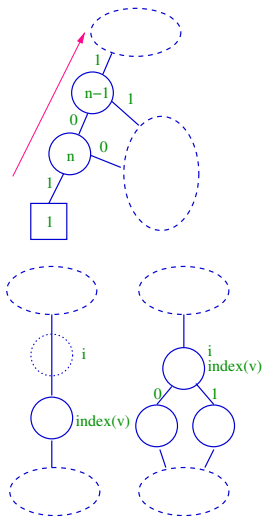


# A Satisfy-one Example



```

procedure Satisfy-all(i: integer; v: vertex; x: array[1..n] of integer):
begin
  if value(v) = 0 then return;
  if i = n + 1 and value(v) = 1
  then begin
    Print element x[1], .. ,x[n];
    return;
  end;
  if index(v)  $\neq$  i
  then begin
    x[i] := 0; Satisfy-all(i + 1, v, x);
    x[i] := 1; Satisfy-all(i + 1, v, x);
  end
  else begin
    x[i] := 0; Satisfy-all(i + 1, low(v), x);
    x[i] := 1; Satisfy-all(i + 1, high(v), x);
  end
end
end;
    
```



# Satisfy-count

- 🌐 The procedure *Satisfy-count* computes a value  $\alpha_v$  to each vertex  $v$  in the graph according to the following recursive formula:
  - ☀️ If  $v$  is a terminal vertex:  $\alpha_v = \text{value}(v)$ .
  - ☀️ If  $v$  is a nonterminal vertex:

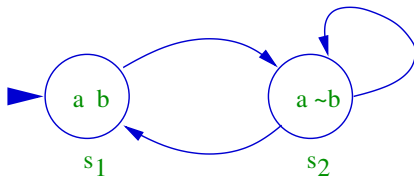
$$\alpha_v = \alpha_{\text{low}(v)} \cdot 2^{\text{index}(\text{low}(v)) - \text{index}(v)} + \alpha_{\text{high}(v)} \cdot 2^{\text{index}(\text{high}(v)) - \text{index}(v)}$$

- 🌐 Once we have computed these values for a graph with root  $v$ , we compute the size of the satisfying set as

$$|S_f| = \alpha_v \cdot 2^{\text{index}(v) - 1}$$

# Kripke Structures

- Given a set of atomic propositions  $AP$ , a Kripke structure  $M$  is a four tuple  $(S, S_0, R, L)$ :
- $S$  is a finite set of states.
  - $S_0 \subseteq S$  is the set of initial states.
  - $R \subseteq S \times S$  is a transition relation that must be total.
  - $L : S \rightarrow 2^{AP}$  is a function that labels each state with the set of atomic propositions true in that state.



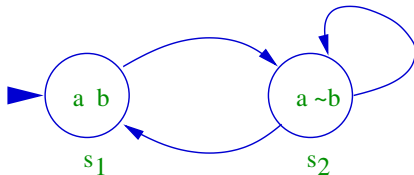
# First Order Representations

- The initial states can be represented by the formula:

$$(a \wedge b)$$

- The transitions can be represented by the formula:

$$\begin{aligned} &(a \wedge b \wedge a' \wedge \neg b') \quad \vee \\ &(a \wedge \neg b \wedge a' \wedge \neg b') \quad \vee \\ &(a \wedge \neg b \wedge a' \wedge b') \end{aligned}$$





# OBDD Representations

- Use  $x_1, x_2, x_3, x_4$  to represent  $a, b, a', b'$  respectively.
- The characteristic function of initial states:

$$(a \wedge b)$$

becomes

$$(x_1 \cdot x_2)$$

# OBDD Representations (cont.)

🌐 The characteristic function of transitions:

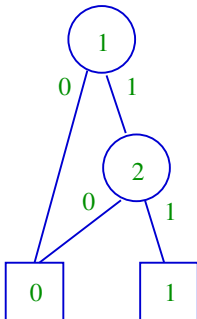
$$\begin{aligned} & (a \wedge b \wedge a' \wedge \neg b') \quad \vee \\ & (a \wedge \neg b \wedge a' \wedge \neg b') \quad \vee \\ & (a \wedge \neg b \wedge a' \wedge b') \end{aligned}$$

becomes

$$\begin{aligned} & (x_1 \cdot x_2 \cdot x_3 \cdot \bar{x}_4) \quad + \\ & (x_1 \cdot \bar{x}_2 \cdot x_3 \cdot \bar{x}_4) \quad + \\ & (x_1 \cdot \bar{x}_2 \cdot x_3 \cdot x_4) \end{aligned}$$

# OBDD Representations (cont.)

Initial states:  $x_1 \cdot x_2$



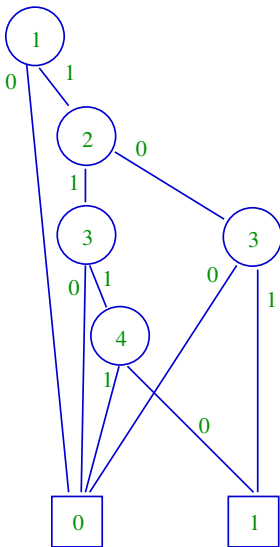
# OBDD Representations (cont.)

Transitions:

$$(x_1 \cdot x_2 \cdot x_3 \cdot \bar{x}_4) \quad +$$

$$(x_1 \cdot \bar{x}_2 \cdot x_3 \cdot \bar{x}_4) \quad +$$

$$(x_1 \cdot \bar{x}_2 \cdot x_3 \cdot x_4)$$



- 🌐 OBDDs are representations of Boolean functions with
  - ☀ canonical forms and
  - ☀ reasonable size.
- 🌐 Transition systems can be encoded in Boolean functions and thus representable in OBDDs.
- 🌐 Symbolic model checking becomes possible with OBDDs.