

# Automated Reasoning

## Lecture 19: Operations on Binary Decision Diagrams (BDDs)

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*based on originals by Paul Jackson  
diagrams from Huth & Ryan, LiCS, 2nd Ed.*

Tuesday 24th March 2015

# Recap

- ▶ Previously:
  - ▶ (Reduced, Ordered) Binary Decision Diagrams ((RO)BDDs)
- ▶ This time:
  - ▶ Operations on ROBDDs
    - reduce, apply, restrict, exists
  - ▶ Symbolic Model Checking with BDDs

# Binary Decision Diagrams

Binary Decision Diagrams: DAGs, such that

- ▶ Unique root node
- ▶ Variables on non-terminal nodes
- ▶ Truth-values on terminal nodes
- ▶ Exactly two edges from each non-terminal node, labelled 0, 1

Some notation, for a given BDD node  $n$ :

- ▶ If  $n$  is a non-terminal node:
  - $\text{var}(n)$  – the variable label on node  $n$ ;
  - $\text{lo}(n)$  – the node reached by following the 0 edge from  $n$ ;
  - $\text{hi}(n)$  – the node reached by following the 1 edge from  $n$ ;
- ▶ If  $n$  is a terminal node:
  - $\text{val}(n)$  – the truth value labelling  $n$

For a BDD  $B$ , the root node is called  $\text{root}(B)$ .

## reduce

`reduce` constructs an ROBDD from an OBDD.

1. Label each BDD node  $n$  with an integer  $\text{id}(n)$ ,
2. in a single bottom-up pass, such that:
3. two BDD nodes  $m$  and  $n$  have the same label ( $\text{id}(m) = \text{id}(n)$ ) if and only if  $m$  and  $n$  represent the same boolean function.

The ROBDD is then created by using one node from each class of nodes with the same label.

## reduce

Assignment of labels follows the rules for performing reductions.

To label a node  $n$ :

- ▶ Remove duplicate terminals:

if  $n$  is a terminal node (*i.e.*,  $\boxed{0}$  or  $\boxed{1}$ ), then set  $\text{id}(n)$  to be  $\text{val}(n)$ .

- ▶ Remove redundant tests:

if  $\text{id}(\text{lo}(n)) = \text{id}(\text{hi}(n))$  then set  $\text{id}(n)$  to be  $\text{id}(\text{lo}(n))$ .

- ▶ Remove duplicate nodes:

if there exists a node  $m$  that has already been labelled such that

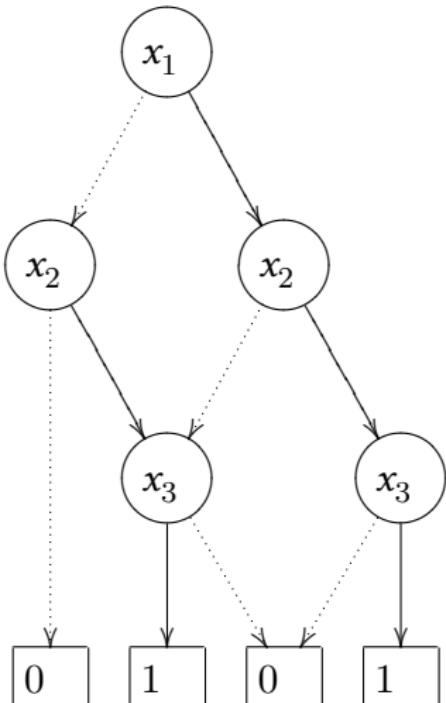
$$\left\{ \begin{array}{l} \text{var}(m) = \text{var}(n) \\ \text{lo}(m) = \text{lo}(n) \\ \text{hi}(m) = \text{hi}(n) \end{array} \right\}, \text{ set } \text{id}(n) \text{ to } \text{id}(m).$$

Use a hashtable with  $\langle \text{var}(n), \text{lo}(n), \text{hi}(n) \rangle$  keys for  $O(1)$  lookup time.

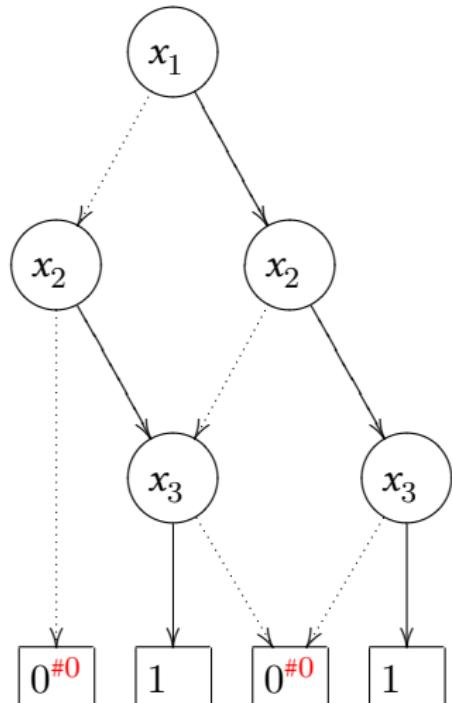
- ▶ Otherwise, set  $\text{id}(n)$  to an unused number.

Using the “big array” approach to storing BDD nodes,  $\text{id}(n)$  is simply the index of the node in the array.

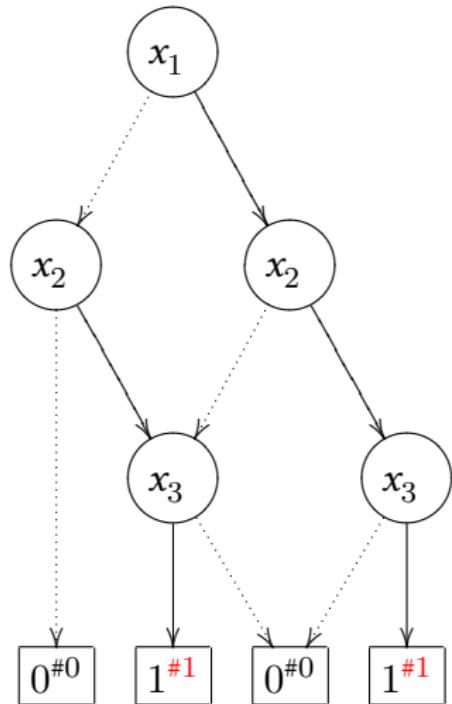
## reduce Example



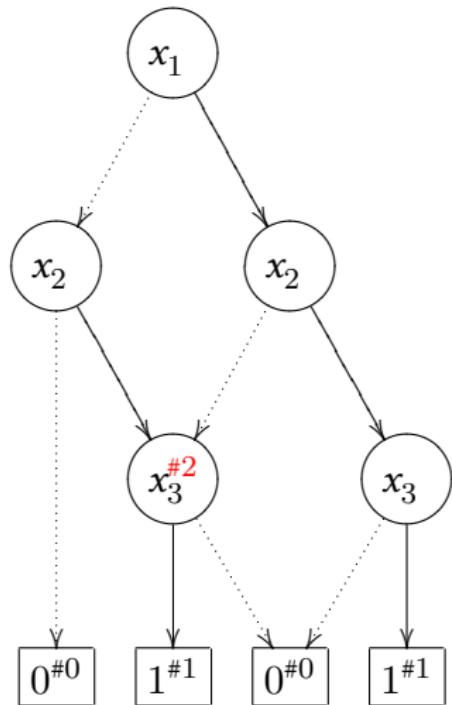
## reduce Example



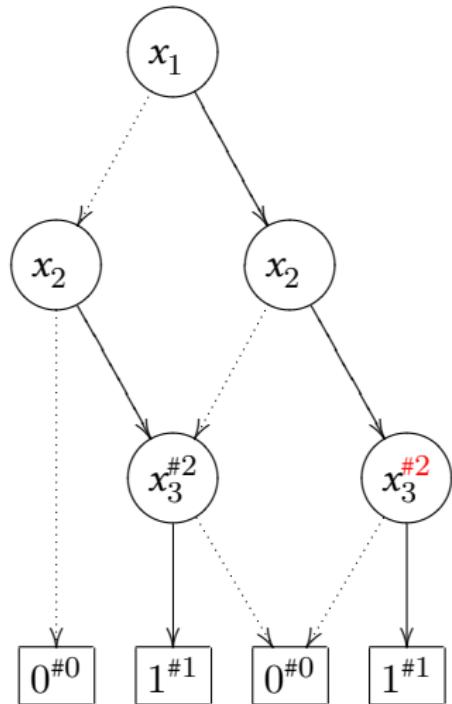
## reduce Example



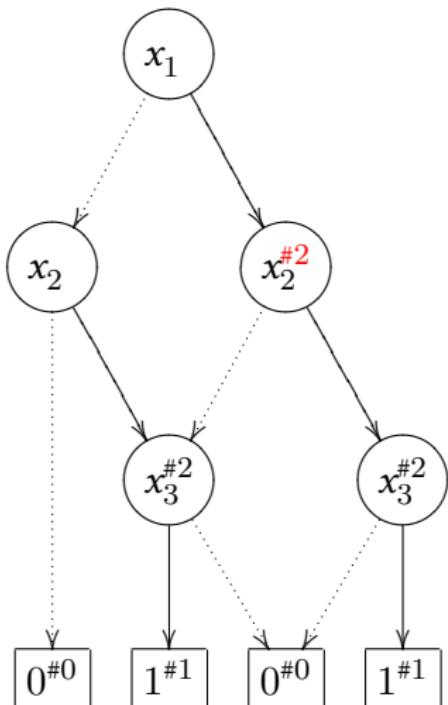
## reduce Example



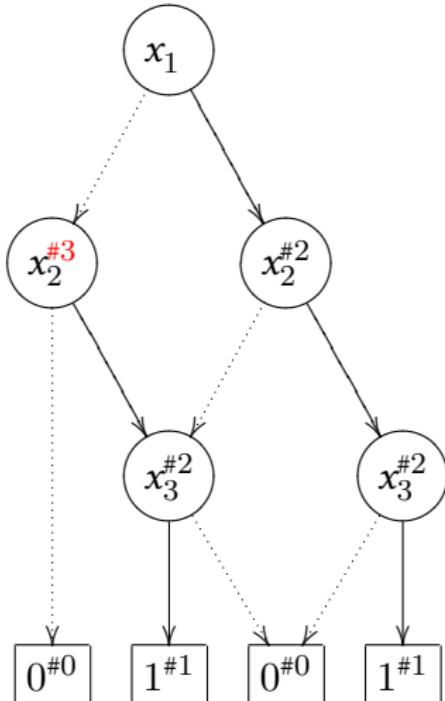
## reduce Example



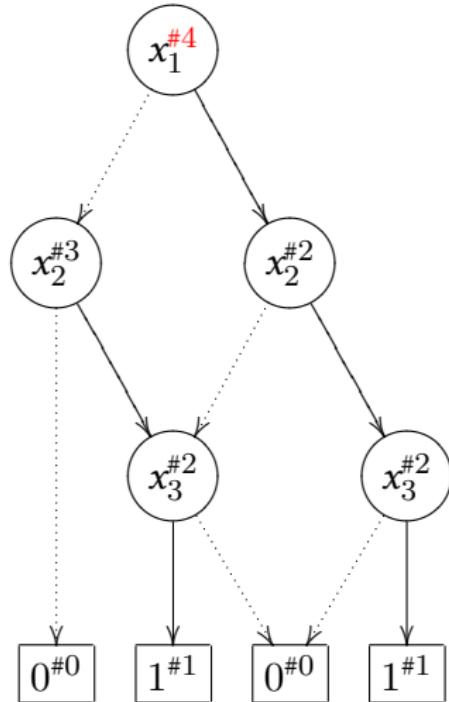
## reduce Example



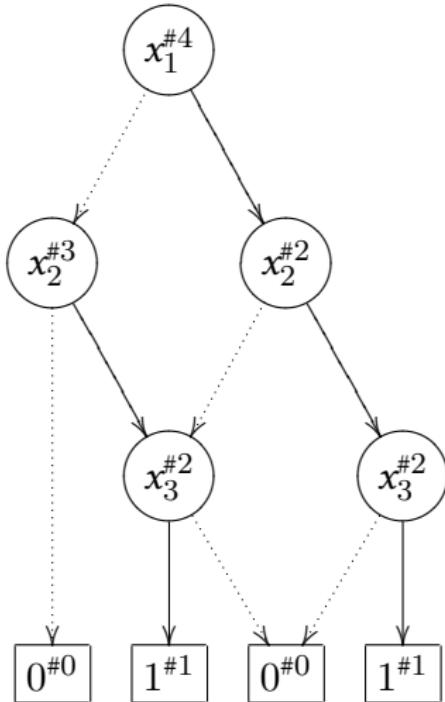
## reduce Example



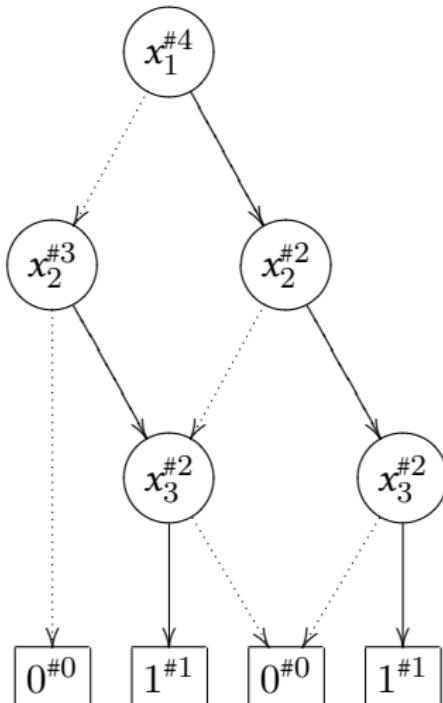
## reduce Example



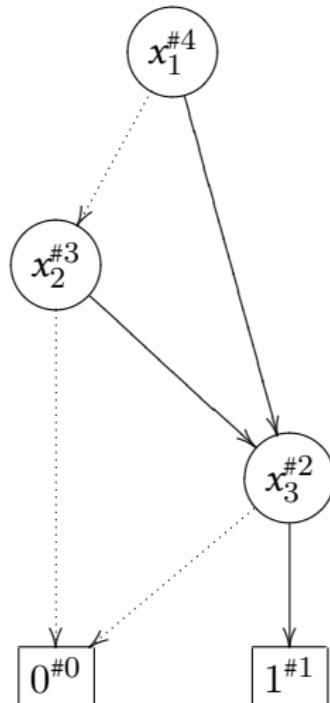
## reduce Example



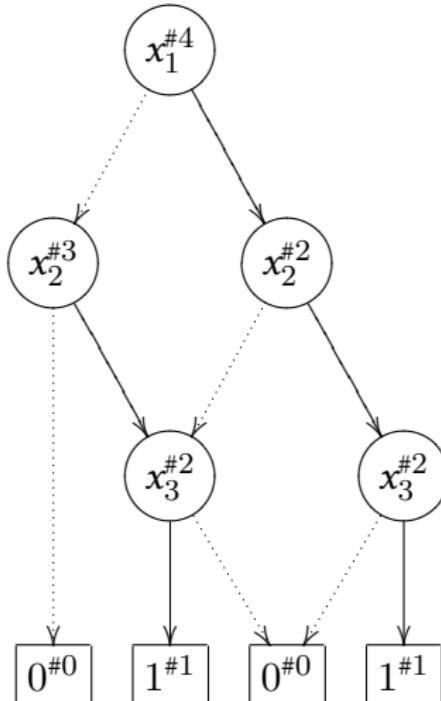
## reduce Example



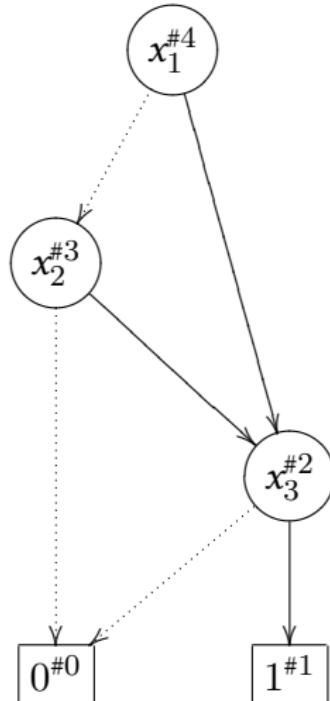
Reduces to



## reduce Example



Reduces to



In practice, labelling and construction are interleaved.

## apply

Given compatible OBDDs  $B_f$  and  $B_g$  that represent formulas  $f$  and  $g$ ,  
 $\text{apply}(\square, B_f, B_g)$  computes a OBDD representing  $f \square g$ .

- ▶ where  $\square$  represents some binary operation on boolean formulas  
*for example,  $\wedge, \vee, \oplus$*
- ▶ Unary operations can be handled too.  
*for example, negation:  $x \square y = x \oplus 1$*

## apply: Shannon expansions

For any boolean formula  $f$  and variable  $x$ , it can be written as:

$$f \equiv (\neg x \wedge f[0/x]) \vee (x \wedge f[1/x])$$

This is the **Shannon expansion** of  $f$ (originally due to G. Boole).

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In particular:  $f \square g$  can be expanded like so:

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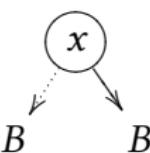
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If a BDD            represents a boolean function  $f$ , then:

1.  $B$  represents  $f[0/x]$  and  $B'$  represents  $f[1/x]$ ; and
2. The BDD is effectively a compressed representation of  $f$  in Shannon normal form.

So: implement apply recursively on the structure of the BDDs.

## apply: cases

$$\text{apply}(\square, \begin{array}{c} x \\ \swarrow \quad \searrow \\ B \quad C \end{array}, \begin{array}{c} x \\ \swarrow \quad \searrow \\ B' \quad C' \end{array}) = \begin{array}{c} x \\ \swarrow \quad \searrow \\ \text{apply}(\square, B, C) \quad \text{apply}(\square, B', C') \end{array}$$

$$\text{apply}(\square, \begin{array}{c} x \\ \swarrow \quad \searrow \\ B \quad B' \end{array}, \begin{array}{c} C \end{array}) = \begin{array}{c} x \\ \swarrow \quad \searrow \\ \text{apply}(\square, B, C) \quad \text{apply}(\square, B', C) \end{array}$$

when  $C$  is terminal node, or non-terminal with  $\text{var}(\text{root}(C)) > x$

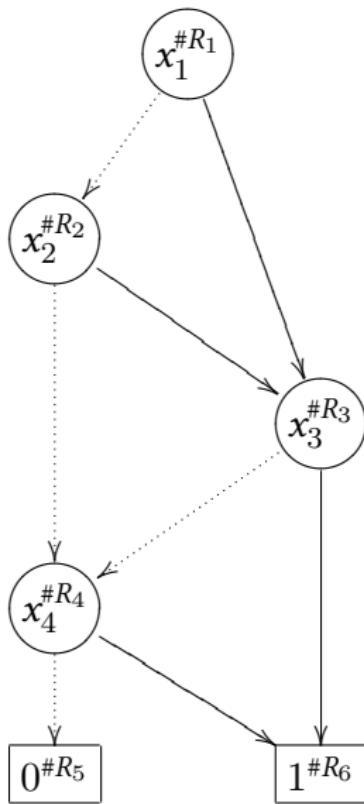
$$\text{apply}(\square, \begin{array}{c} B \\ , \end{array}, \begin{array}{c} x \\ \swarrow \quad \searrow \\ C \quad C' \end{array}) = \begin{array}{c} x \\ \swarrow \quad \searrow \\ \text{apply}(\square, B, C) \quad \text{apply}(\square, B, C') \end{array}$$

when  $B$  is terminal node, or non-terminal with  $\text{var}(\text{root}(B)) > x$

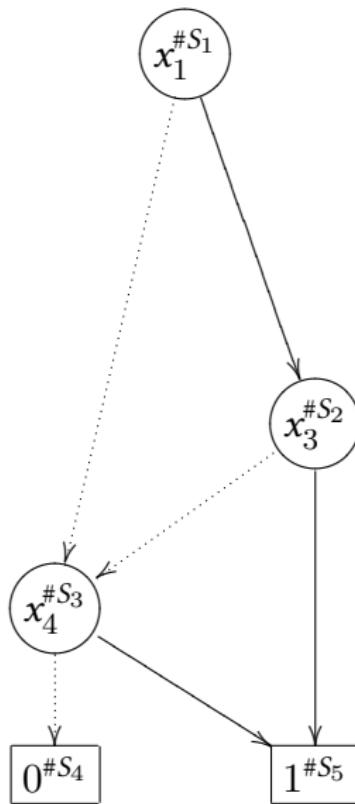
$$\text{apply}(\square, \boxed{u}, \boxed{v}) = \boxed{u \square v}$$

## apply: example

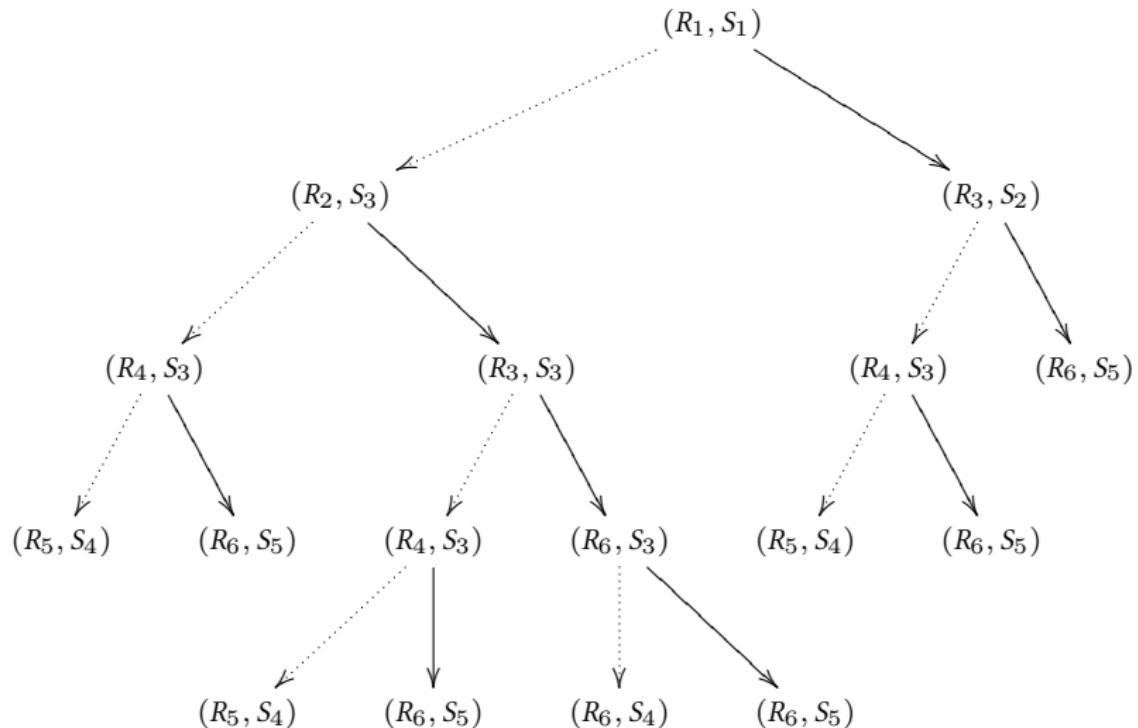
Compute  $\text{apply}(\vee, B_f, B_g)$ , where  $B_f$  and  $B_g$  are:



$\vee$



## apply: recursive calls



## apply: memoisation

The recursive apply implementation will generate an OBBD.

- ▶ Apply reduce to convert it back to an ROBDD.

However, as can be seen from the tree of recursive calls, there are many calls to apply with the same arguments.

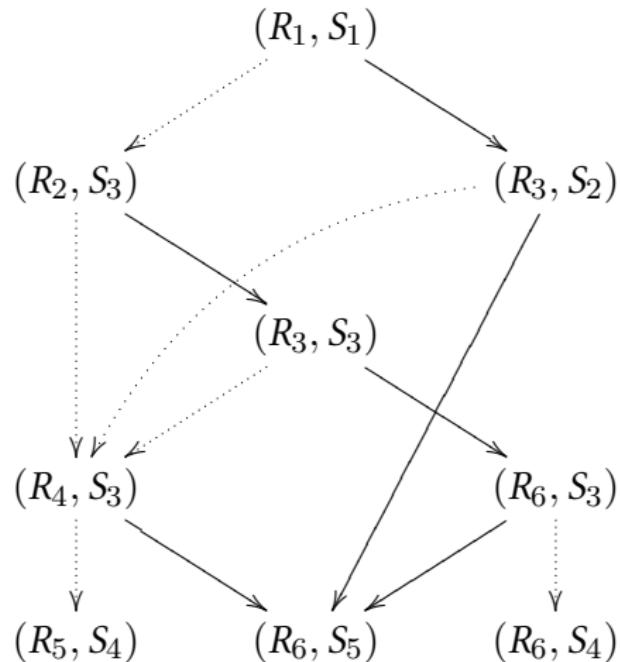
- ▶ Each invocation of apply where at least one of the arguments is non-terminal generates two further calls to apply: the number of calls is worst-case exponential in the sizes of the original diagrams.

We are not taking into account the **sharing** in BDDs.

We can greatly improve the run-time by using **memoisation**: remembering the results of previous calls.

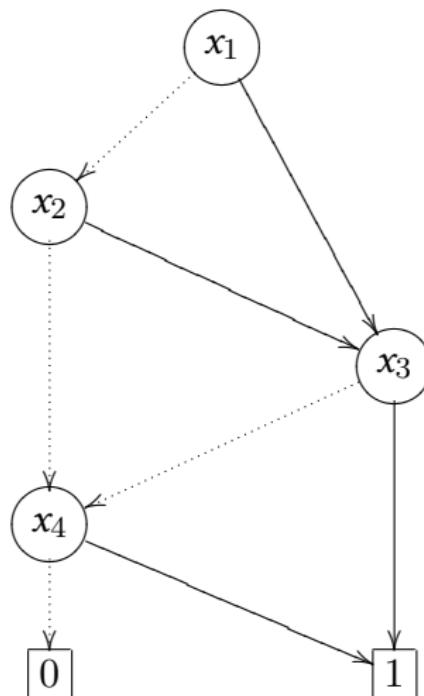
## apply: memoised recursive calls

Memoisation results in at most  $|B_f| \cdot |B_g|$  calls to apply.



## apply: Result

If we are careful to never create the same BDD node twice (using the same lookup table technique as `reduce`), then with memoisation, we automatically get a reduced BDD:



## Other Operations

`restrict(0, x, Bf)` computes ROBDD for  $f[0/x]$

1. For each node  $n$  labelled with  $x$ , incoming edges are redirected to  $\text{lo}(n)$ , and the node  $n$  is removed.
2. Resulting BDD then reduced with `reduce`.
3. (again, `reduce` can be interleaved with the removal.)

`exists(x, Bf)` computes ROBDD for  $\exists x. f$ .

1. Uses the identity

$$(\exists x. f) \equiv f[0/x] \vee f[1/x]$$

2. Realised using the `restrict` and `apply` functions:

`apply( $\vee$ , restrict(0, x, Bf), restrict(1, x, Bf))`

# Time Complexities

Algorithm	Input OBDDs	Output OBDD	Time complexity
reduce	$B$	reduced $B$	$O( B  \cdot \log  B )$
apply	$B_f, B_g$ (reduced)	$B_{f \square g}$ (reduced)	$O( B_f  \cdot  B_g )$
restrict	$B_f$ (reduced)	$B_{f[0/x]}$ or $B_{f[1/x]}$ (red'd)	$O( B_f  \cdot \log  B_f )$
$\exists$	$B_f$ (reduced)	$B_{\exists x_1 \dots x_n.f}$ (reduced)	NP-complete

H&R, Figure 6.23

# Implementing CTL Model Checking using BDDs

Recall:

1. CTL model checking computes a set of states  $\llbracket \phi \rrbracket$  for every sub-formula  $\phi$  of the original formula.
2. Sets of states will be represented using ROBDDs

States are represented by boolean vectors  $\langle v_1, \dots, v_n \rangle$ .

Sets of states are represented using ROBDDs on  $n$  variables  $x_1, \dots, x_n$  that describe the characteristic function of the set.

- ▶ Operations on sets are implemented using the operations on BBDs

For example, the definition

$$\llbracket \phi \wedge \psi \rrbracket = \llbracket \phi \rrbracket \cap \llbracket \psi \rrbracket$$

Is implemented by:

$$B_{\llbracket \phi \wedge \psi \rrbracket} = \text{apply}(\wedge, B_{\llbracket \phi \rrbracket}, B_{\llbracket \psi \rrbracket})$$

# Implementing CTL Model Checking using BDDs

Transition relations ( $\rightarrow$ )  $\subseteq S \times S$  are represented by ROBDDs on  $2n$  variables.

- If the variables  $x_1, \dots, x_n$  describe the current state, and the variables  $x'_1, \dots, x_n$  describe the next state, then a good ordering is  $x_1, x'_1, x_2, x'_2, \dots, x_n, x'_n$  (interleaving).

When translating from the model description, the boolean formulas describing the:

1. initial state set
2. transition relation
3. defined variables

are translated into ROBDDs by using the apply algorithm, following the structure of the original formula.

This avoids exponential blow-up from first constructing a decision tree and then reducing.

# Implementing CTL Model Checking using BDDs

The function applications

$$\begin{aligned}\text{pre}_{\exists}(Y) & \doteq \{s \in S \mid \exists s' \in S. (s \rightarrow s') \wedge s' \in Y\} \\ \text{pre}_{\forall}(Y) & \doteq \{s \in S \mid \forall s' \in S. (s \rightarrow s') \rightarrow s' \in Y\}\end{aligned}$$

are implemented using BDDs like so:

$$B_{\text{pre}_{\exists}(Y)} = \text{exists}(\vec{x}', \text{apply}(\wedge, B_{\rightarrow}, B_{Y'}))$$

where

- ▶  $B_{\rightarrow}$  is the ROBDD representing the transition relation  $\rightarrow$ ;
- ▶  $B_{Y'}$  is the RODBB representing the set  $Y$  with the variables  $x_1, \dots, x_n$  renamed to  $x'_1, \dots, x'_n$ .

And:

$$\text{pre}_{\forall}(Y) = S - \text{pre}_{\exists}(S - Y)$$

where  $S - Y$  is implemented by negation (via apply).

# Implementing CTL Model Checking using BDDs

To implement the temporal connectives, we compute fix points.

$$\begin{aligned}\llbracket \text{EF } \phi \rrbracket &= \mu Y. \llbracket \phi \rrbracket \cup \text{pre}_{\exists}(Y) \\ \llbracket \text{EG } \phi \rrbracket &= \nu Y. \llbracket \phi \rrbracket \cap \text{pre}_{\exists}(Y) \\ &\dots\end{aligned}$$

By Knaster-Tarski, we know that:

- ▶  $F^{|S|}(\emptyset)$  is the *least* fixed point of  $F$ :  $\mu Y.F(Y)$
- ▶  $F^{|S|}(S)$  is the *greatest* fixed point of  $F$ :  $\nu Y.F(Y)$

Compute  $\llbracket \text{EF } \phi \rrbracket$  using the sequence (of ROBDDs)

$$Y^0 = \emptyset, Y^1 = \llbracket \phi \rrbracket \cup \text{pre}_{\exists}(\emptyset), Y^2 = \llbracket \phi \rrbracket \cup \text{pre}_{\exists}(\llbracket \phi \rrbracket \cup \text{pre}_{\exists}(\emptyset)), \dots$$

Usually, we won't need  $|S|$  steps: we can stop when  $Y_i = Y_{i+1}$

- ▶ This check is very cheap with ROBDDs.

# Summary

- ▶ Operations on BDDs (H&R 6.2)
  - ▶ reduce
  - ▶ apply
  - ▶ restrict, exists
- ▶ Symbolic Model Checking (H&R 6.3)
  - ▶ Representing states and transitions as BDDs
  - ▶ Implementing the CTL MC algorithm with BDDs
- ▶ Next:
  - ▶ Friday 27th March: Phil Scott  
*“Formalising the Foundations of Geometry”*
  - ▶ Next Tuesday (31st March): Exam Review