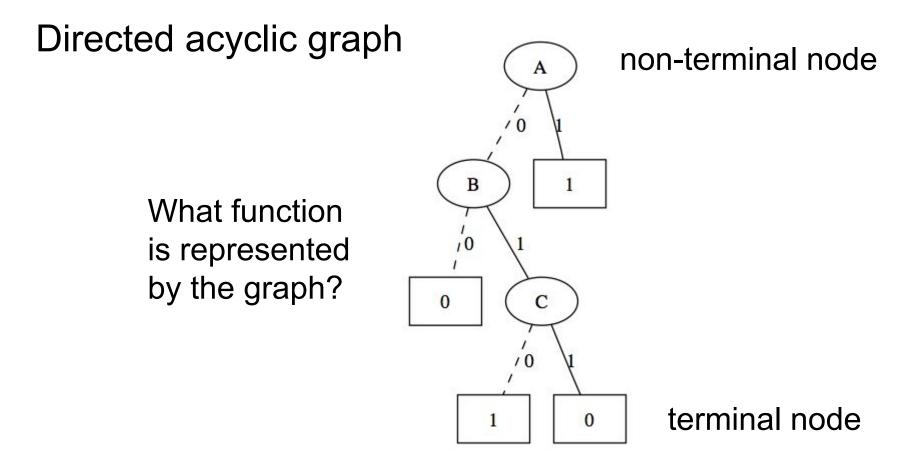
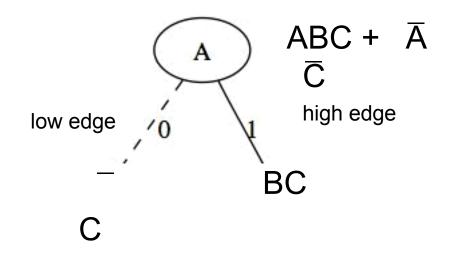
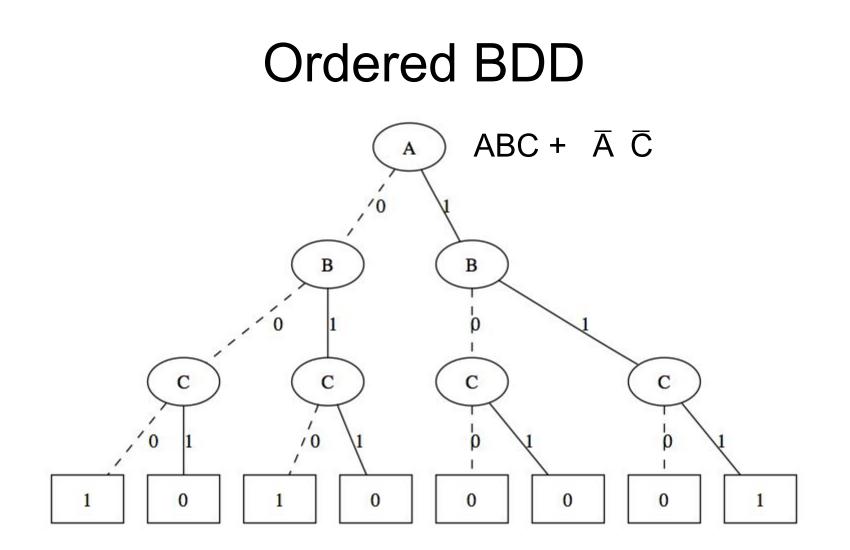
Binary Decision Diagrams



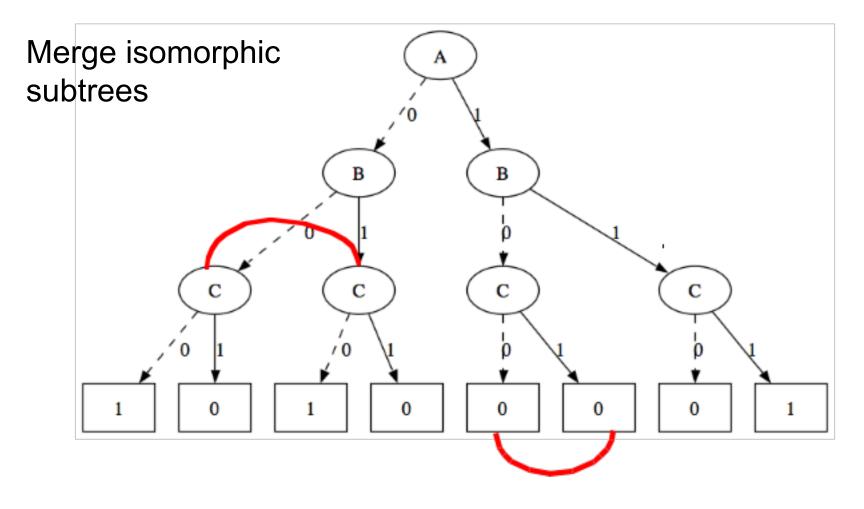
Shannon Decomposition



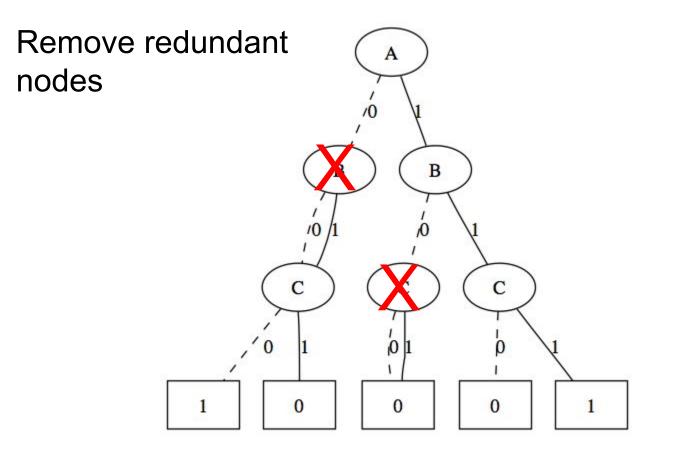
Apply the decomposition recursively



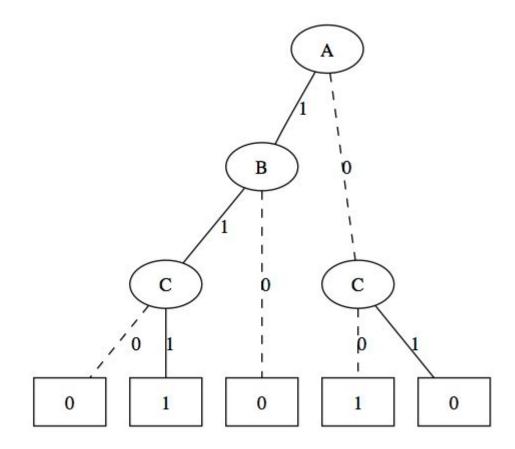
Reduced Ordered BDD



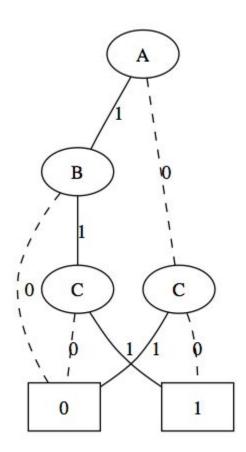
Reduced Ordered BDD



ROBDD



ROBDD



BDD

A Binary Decision Diagram (BDD) is a rooted, directed acyclic graph

- with one or two terminal nodes of out-degree zero labeled 0 or 1, and a set of variable nodes u of out-degree two.
- The two outgoing edges are given by two functions low(u) and high(u). (In pictures, these are shown as dotted and solid lines, respectively.) A variable var(u) is associated with each variable node.

ROBDD

- A BDD is Ordered (OBDD) if on all paths through the graph the variables respect a given linear order $x_1 < x_2 < \ldots < x_n$. An OBDD is Reduced (ROBDD) if
- 1. (**uniqueness**) no two distinct nodes u and v have the same variable name and low- and highsuccessor, i.e.,

var(u) = var(v); low(u) = low(v); high(u) = high(v) implies u = v;

and

2. (non-redundant tests) no variable node u has identical low- and high-successor, i.e., low(u) = high(u)

Canonicity

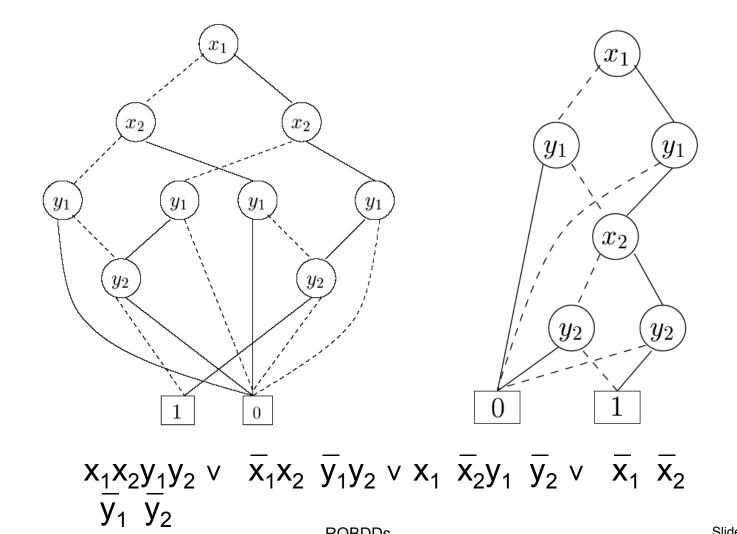
For any function f : Bⁿ →B there is exactly one ROBDD u with variable ordering x₁ < x₂ < ... < x_n such that f^u = f(x₁, ..., x_n).
 Proof:

Consequences of Canonicity

- How do you represent a tautology? (i.e. all variable assignments yield 1)
- How do you know that the function is satisfiable?

(i.e. there is at least one assignment for the variables such that the function evaluates to 1)

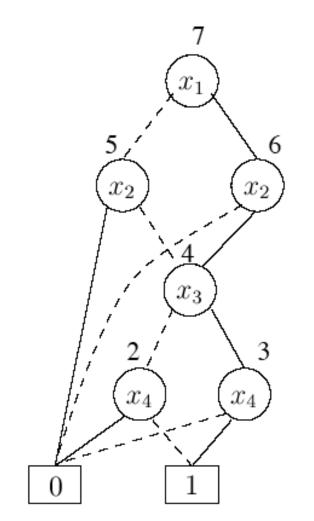
Variable Order



ROBDDs

Constructing and manipulating BDDs

- Nodes will be represented as numbers 0, 1, 2, ... with 0 and 1 reserved for the terminal nodes.
- The variables in the ordering $x_1 < x_2 < ... < x_n$ are represented by their indices 1, 2, ..., n.
- The ROBDD is stored in a table T:u →(i, l, h) which maps a node u to its three attributes var(u) = i, low(u) = l, and high(u) = h.



$T: u \mapsto (i, l, h)$			
u	var	low	high
0	5		
1	5		
2	4	1	0
3	4	0	1
4	3	2	3
5	2	4	0
6	2	0	4
7	1	5	6

The H Table

 In order to ensure that the OBDD being constructed is reduced, it is necessary to determine from a triple (i, I, h) whether there exists a node u with var(u) = i, low(u) = l, and high(u) = h. For this purpose we assume the presence of a table H : (i, I, h) \rightarrow u mapping triples (i, l, h) of variable indices i, and nodes I, h to nodes u. The table H is the "inverse" of the table T, i.e., for variable nodes u, T(u) = (i, i)I, h), if and only if, H(i, I, h) = u.

Operations on T and H

$$T: u \mapsto (i, l, h)$$

$$init(T)$$

$$u \leftarrow add(T, i, l, h)$$

$$var(u), low(u), high(u)$$

initialize T to contain only 0 and 1 allocate a new node u with attributes (i, l, h)lookup the attributes of u in T

$$\begin{array}{ll} H:(i,l,h)\mapsto u\\ init(H) & \text{initialize } H \text{ to be empty}\\ b\leftarrow member(H,i,l,h) & \text{check if } (i,l,h) \text{ is in } H\\ u\leftarrow lookup(H,i,l,h) & \text{find } H(i,l,h)\\ insert(H,i,l,h,u) & \text{make } (i,l,h) \text{ map to } u \text{ in } H \end{array}$$

The Function Mk

$M\kappa[T, H](i, l, h)$ 1: if l = h then return l2: else if member(H, i, l, h) then 3: return lookup(H, i, l, h)4: else $u \leftarrow add(T, i, l, h)$ 5: insert(H, i, l, h, u)6: return u

What is the complexity of Mk?

The Build Function

```
BUILD[T, H](t)
     function BUILD'(t, i) =
1:
           if i > n then
2:
                  if t is false then return 0 else return 1
3:
            else v_0 \leftarrow \text{BUILD'}(t[0/x_i], i+1)
4:
                  v_1 \leftarrow \text{BUILD'}(t[1/x_i], i+1)
5:
                  return MK(i, v_0, v_1)
6:
7:
      end BUILD'
8:
9:
      return BUILD'(t, 1)
```

• Show build $Build(A \oplus B \oplus C)$

- What is the running time of Build?
- Can it be improved?
- How?

Apply

All the binary Boolean operators on ROBDDs are implemented by the same general algorithm APPLY(op, u_1, u_2) that for two ROBDDs computes the ROBDD for the Boolean expression t^{u_1} op t^{u_2} . The construction of APPLY is based on the Shannon expansion (2):

$$t = x \to t[1/x], t[0/x].$$

Observe that for all Boolean operators op the following holds:

$$(x \to t_1, t_2) \ op \ (x \to t'_1, t'_2) \ = \ x \to t_1 \ op \ t'_1, t_2 \ op \ t'_2 \tag{4}$$

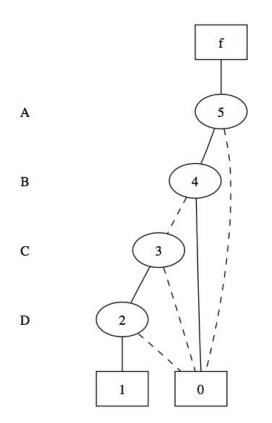
The Function Apply

```
\operatorname{APPLY}[T, H](op, u_1, u_2)
1:
2:
    function APP(u_1, u_2) =
3:
4:
           if u_1 \in \{0, 1\} and u_2 \in \{0, 1\} then u \leftarrow op(u_1, u_2)
5:
      else if var(u_1) = var(u_2) then
6:
            u \leftarrow MK(var(u_1), APP(low(u_1), low(u_2)), APP(high(u_1), high(u_2)))
7:
      else if var(u_1) < var(u_2) then
8
9
            u \leftarrow MK(var(u_1), APP(low(u_1), u_2), APP(high(u_1), u_2))
     else (* var(u_1) > var(u_2) *)
10:
            u \leftarrow MK(var(u_2), APP(u_1, low(u_2)), APP(u_1, high(u_2)))
11:
12:
13:
     return u
                                               Complexity?
14: end APP
15:
16: return APP(u_1, u_2)
```

The Function Apply

```
\operatorname{APPLY}[T, H](op, u_1, u_2)
                                     Dynamic programming
   init(G)
1:
2:
   function APP(u_1, u_2) =
3:
     if G(u_1, u_2) \neq empty then return G(u_1, u_2)
4:
     else if u_1 \in \{0, 1\} and u_2 \in \{0, 1\} then u \leftarrow op(u_1, u_2)
5:
     else if var(u_1) = var(u_2) then
6:
           u \leftarrow MK(var(u_1), APP(low(u_1), low(u_2)), APP(high(u_1), high(u_2)))
7:
     else if var(u_1) < var(u_2) then
8
9
           u \leftarrow MK(var(u_1), APP(low(u_1), u_2), APP(high(u_1), u_2))
     else (* var(u_1) > var(u_2) *)
10:
           u \leftarrow MK(var(u_2), APP(u_1, low(u_2)), APP(u_1, high(u_2)))
11:
     G(u_1, u_2) \leftarrow u
12:
13:
     return u
                                              Complexity?
14: end APP
15:
16: return APP(u_1, u_2)
```

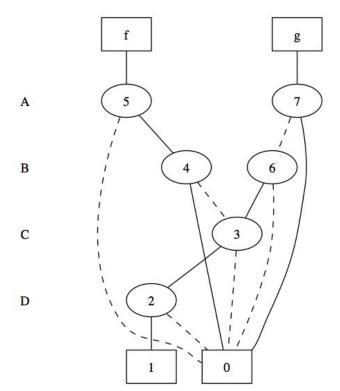
$f = A \overline{B}CD$



i	v	Ι	h
0	5		
1	5		
2	4	0	1
3	3	0	2
4	2	3	0
5	1	0	4

add function $g = \overline{A}BCD$

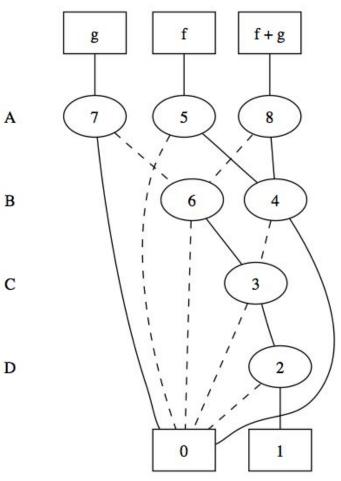
 $f = A \overline{B}CD$ $g = \overline{A}BCD$



i	v	I	h
0	5		
1	5		
2	4	0	1
3	3	0	2
4	2	3	0
5	1	0	4
6	2	0	3
7	1	6	0

APPLY(f,g,+)

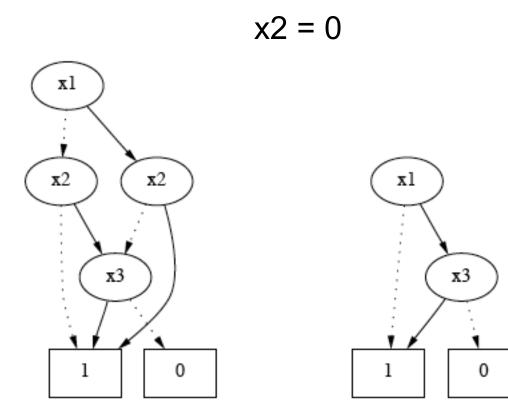
i	V	I	h	
0	5			
1	5			
2	4 3	0	1	1
2 3 4 5	3	0	2	
4	2	3	0	
	1	0	4 3	
6	2	0	3	
7	1	6	0]
8	1	6	4	



This is a forest (many trees)

How many nodes can a forest have?

Restrict



Restrict

RESTRICT[T, H](u, j, b) =1: function res(u) =2: if var(u) > j then return u3: else if var(u) < j then return MK(var(u), res(low(u)), res(high(u)))4: else (* var(u) = j *) if b = 0 then return res(low(u))5: else (* var(u) = j, b = 1 *) return res(high(u))6: end res 7: return res(u)

SatCount

• Count the number of assignment for which the function is true

- 5: return res
- 6: end count
- 7:

8: return
$$2^{var(u)-1} * count(u)$$

AnySat

Find an assignment for which the function is true

 $\operatorname{AnySat}(u)$

- 1: **if** u = 0 **then** Error
- 2: else if u = 1 then return []
- 3: else if low(u) = 0 then return $[x_{var(u)} \mapsto 1, ANYSAT(high(u))]$
- 4: else return $[x_{var(u)} \mapsto 0, ANYSAT(low(u))]$

AllSat

• Find all assignments for which the function is true

AllSat(u)
1:	if $u = 0$ then return $\langle \rangle$
2:	else if $u = 1$ then return $\langle [] \rangle$
3:	else return
4:	$(\text{add } [x_{var(u)} \mapsto 0] \text{ in front of all }$
5:	truth-assignments in $ALLSAT(low(u))$,
6:	add $[x_{var(u)} \mapsto 1]$ in front of all
7:	truth-assignments in $ALLSAT(high(u))$

Simplify

```
SIMPLIFY(d, u)
1:
     function sim(d, u)
          if d = 0 then return 0
2:
3:
          else if u \leq 1 then return u
          else if d = 1 then
4:
               return MK(var(u), sim(d, low(u)), sim(d, high(u)))
5:
          else if var(d) = var(u) then
6:
               if low(d) = 0 then return sim(high(d), high(u))
7:
               else if high(d) = 0 then return sim(low(d), low(u))
8:
               else return MK(var(u),
9:
10:
                                 sim(low(d), low(u)),
                                sim(high(d), high(u)))
11:
          else if var(d) < var(u) then
12:
               return MK(var(d), sim(low(d), u), sim(high(d), u))
13:
14:
          else
15:
               return MK(var(u), sim(d, low(u)), sim(d, high(u)))
16:
     end sim
17:
18:
     return sim(d, u)
```

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Optimizations

- deleted nodes
 - memory management
- negated edges
- variable reordering
 - sifting