


Boolean Function Representation and Manipulation

Ordinary Binary Decision Diagram

Chapter 6

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BDD – Based Upon Shannon Expansion

- Notations


- ◆ $f(x_1, x_2, \dots, x_n)$ - n-input function, $x_i = 0$ or 1

- ◆ $f_{|x_i=b}(x_1, \dots, x_n) = f(x_1, \dots, x_{i-1}, b, x_{i+1}, \dots, x_n)$, $b=0$ or 1

- Shannon Expansion – fix a variable

- ◆ $f = x_i f_{|x_i=1}(x_1, \dots, x_n) + x_i' f_{|x_i=0}(x_1, \dots, x_n)$

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OBDD

- From root (x1) to a terminal vertex (0 or 1), the ordering of *the splitting variables* is the same
- Each sub-function appears only once
 - ◆ two sub-functions are the same iff their OBDD graphs are the same \Leftrightarrow canonical representation

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Classical OBDD Example

- $2n+1$ vertices
- Variable ordering is not important

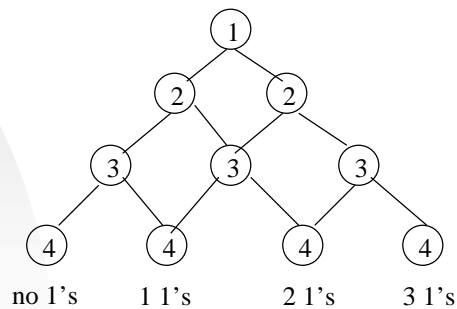
$f = x_1 \oplus x_2 \oplus \dots \oplus x_n$

even # of 1's odd # of 1's

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Classical OBDD Example

- To represent a symmetric function
 - ◆ variable ordering is not important
 - ◆ at level i , we need at most i vertices
 - ◆ total nodes is $< O(n^2)$



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Ordering Example

- $f = x_1 x_2 + x_3 x_4 + x_5 x_6$
 - ◆ ordering 1 : $x_1 x_2 x_3 x_4 x_5 x_6$ (linear result)
 - ✦ If $(x_1 x_2) = 1$, $f = 1 \Rightarrow$ no need to read further!
 - ✦ Otherwise we can "forget about $x_1 x_2$ "
 - ✦ The "*memoryless effect*" makes BDD small
 - A node in a BDD is a memory cell to remember the intermediate results during the computation of the function
 - ◆ ordering 2 : $x_1 x_3 x_5 x_2 x_4 x_6$ (exponential result)
 - ✦ If $(x_1=x_3=x_5=1) \Rightarrow$ we need to remember them all!
 - ✦ Because nothing can be decided based upon the information we have

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Ordering and OBDD Size

- A node in a function graph is a place to remember important information about the computation
- $f = x_1 x_2 + x_3 x_4 + \dots + x_{(n-1)} x_n$
 - ◆ *proceed in this direction* →
- when reach x_2
 - ◆ need to remember the value of x_1
- when reach x_3
 - ◆ only when $x_1 x_2 = 0$
 - ◆ if $x_1 x_2 = 0$, then $x_1 x_2$ can be ignored and from x_3 the computation restarts *fresh*

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OBDD Size

- $f = x_1 x_{n/2+1} + x_2 x_{n/2+2} + \dots + x_{n/2} x_n$
- proceed ordering as $x_1 x_2 x_3 x_4 \dots x_{(n-1)} x_n$
- when reach $x_{n/2}$
 - ◆ all values in $x_1 x_2 x_3 \dots x_{(n-1)/2}$ will need to be remembered because a different assignment may lead to different result
 - ◆ nothing can be ignored and nothing for sure to be said about the evaluation yet
 - ◆ need a complete binary tree for $(n-1)/2$ variables
 - ✦ exponential result


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Worst-Case Example

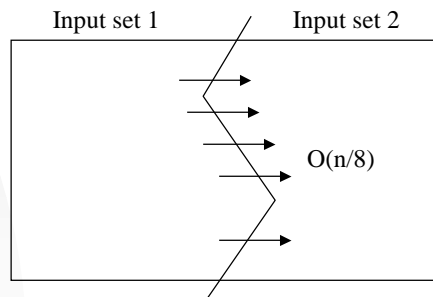
- Worst-case example - multiplication
 - ◆ $f = A * B$
 - ◆ for any input ordering, OBDD(f) requires an exponential number of nodes
- $(a_n a_{n-1} \dots a_1) * (b_n b_{n-1} \dots b_1)$
 - ◆ π is an ordering
 - ◆ Theorem: for any π , OBDD size $\geq 2^{n/8}$
- Potential solutions
 - ◆ word-level BDD
 - ◆ Binary moment diagram (BMD)

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Idea of Proof

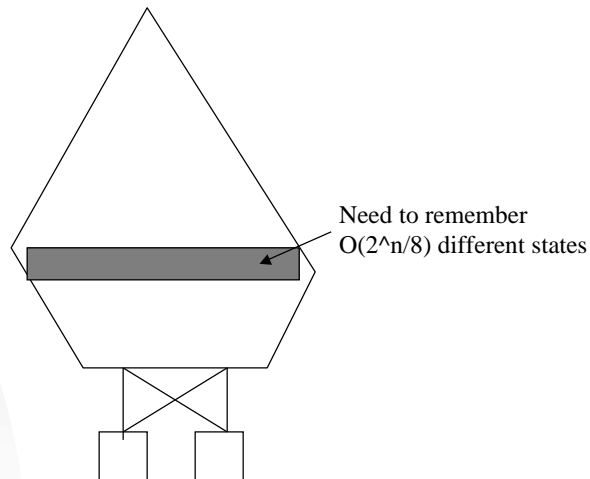
- No matter how you order the input variables, you can always find a cut that
 - ◆ The width is $O(n)$
 - ◆ The information flow is $O(n/8)$



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Equivalent to Say



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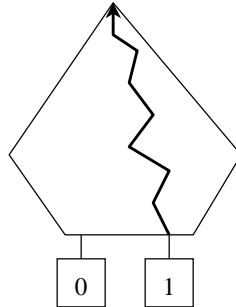
More Terminology

- Composition
 - ◆ $f(x_i=g)(x_1 \dots x_n) = f(x_1, \dots, g(x_1 \dots x_n), \dots, x_n)$
 - ◆ ex. $f=x_1x_2+x_3x_4$ and $g=x_2+x_3$
 - ◆ $f(x_1=g) = x_2 + x_2x_3 + x_3x_4 = x_2 + x_3x_4$
- Dependency Set
 - ◆ $I_f = \{ i \mid f(x_i=0) \neq f(x_i=1) \} \Rightarrow f$ depends on x_i
 - ◆ $f = x_1 x_2 x_3 + x_1' x_2 x_3 \Rightarrow f$ is independent of x_1
- Satisfying Set
 - ◆ $S_f = \{ (x_1 \dots x_n) \mid f(x_1 \dots x_n) = 1 \}$
 - ◆ all assignments that made $f = 1$

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Satisfiability

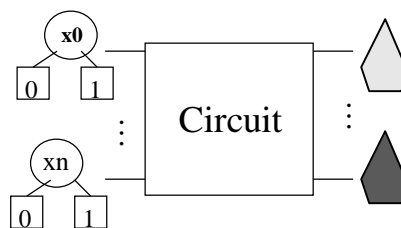


- Satisfy-one: find a path from 1 to the root
- Satisfy-all : find all different paths from 1 to the root
- Satisfy-count : count the # of satisfying assignments

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Computing OBDDs



- Initialize all Pis
- Compute the result of a boolean operation as merging two OBDDs
 - ◆ and, or, xor, etc.
 - ◆ Use procedures *Apply* and *Reduce*
- Follow topological order until reaching all POs


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OBDD Operation Complexity

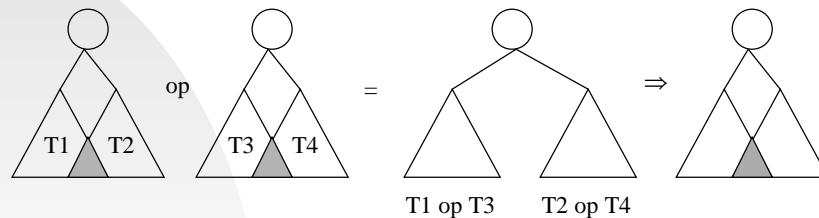
- Reduce (make canonical, minimal form) - $O(G \log G)$
 - ◆ (DA) graphs isomorphism
- Apply (merge two OBDDs) - $O(G_1 G_2)$
 - ◆ Boolean operation = {Apply \Rightarrow Reduce}
- Restrict $\{f(x_i=b)\}$ - $O(G \log G)$
- Compose $\{f(x_i=g)\}$ - $O(G_1^2 G_2)$
- Satisfy-One - $O(n)$
- Satisfy-all - $O(n |S_f|)$
- Satisfy-Count - $O(G)$
 - ◆ This is a #P-complete problem

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Apply

- $f_1 \text{ op } f_2 =$
 - ◆ $x_i' [f_1(x_i=0) \text{ op } f_2(x_i=0)] + x_i [f_1(x_i=1) \text{ op } f_2(x_i=1)]$



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Example

NAND =

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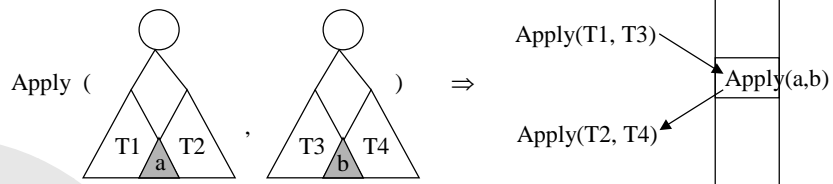
Complexity of Apply

op =

- $k \leq i * j$
 - ◆ i nodes represent i different sub-functions
 - ◆ the worst case - (each of i nodes) op (each of j nodes) are different
 - ◆ Total complexity $O(G1 G2)$

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Hash Table in Apply



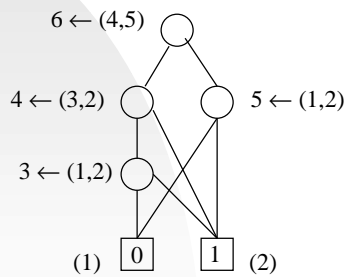
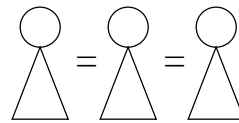
- When doing $\text{Apply}(T1, T3)$, we have done $\text{Apply}(a,b)$
 - ◆ $\text{Apply}(a,b)$ should not be repeated
- Need a hash table to store results of from Apply of all sub-functions.

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Reduce

- DA Graph Isomorphism
 - ◆ we can do a pair-wise comparison for all nodes
 - ◆ but that is too slow
 - ✦ instead: by labeling bottom-up



Each graph has a unique labeling scheme
 \Rightarrow same label at the root = same graph

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Labeling Scheme

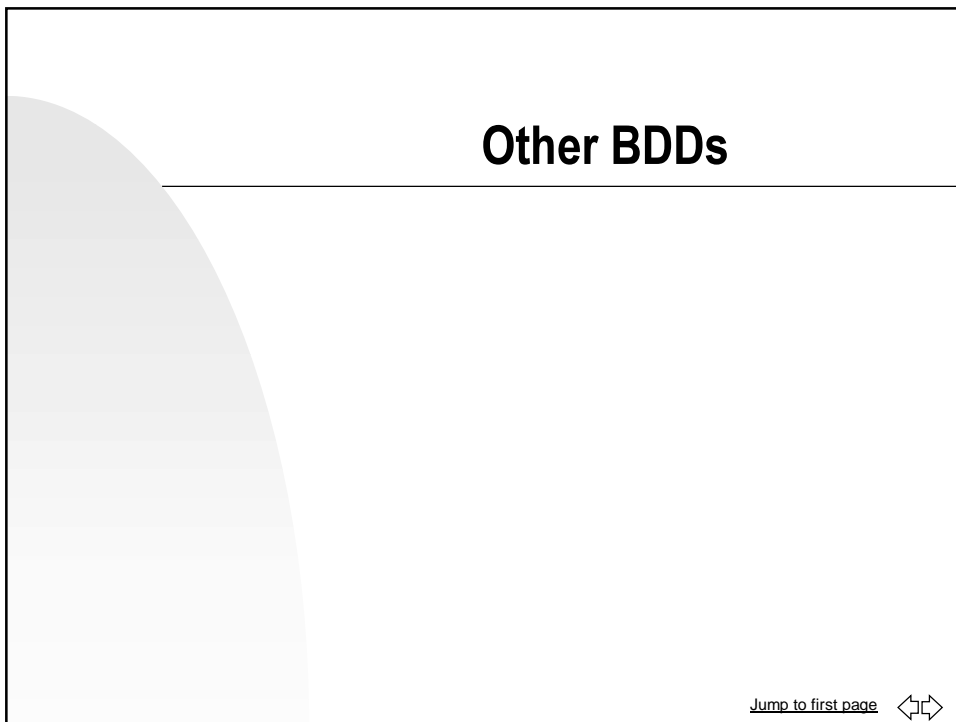
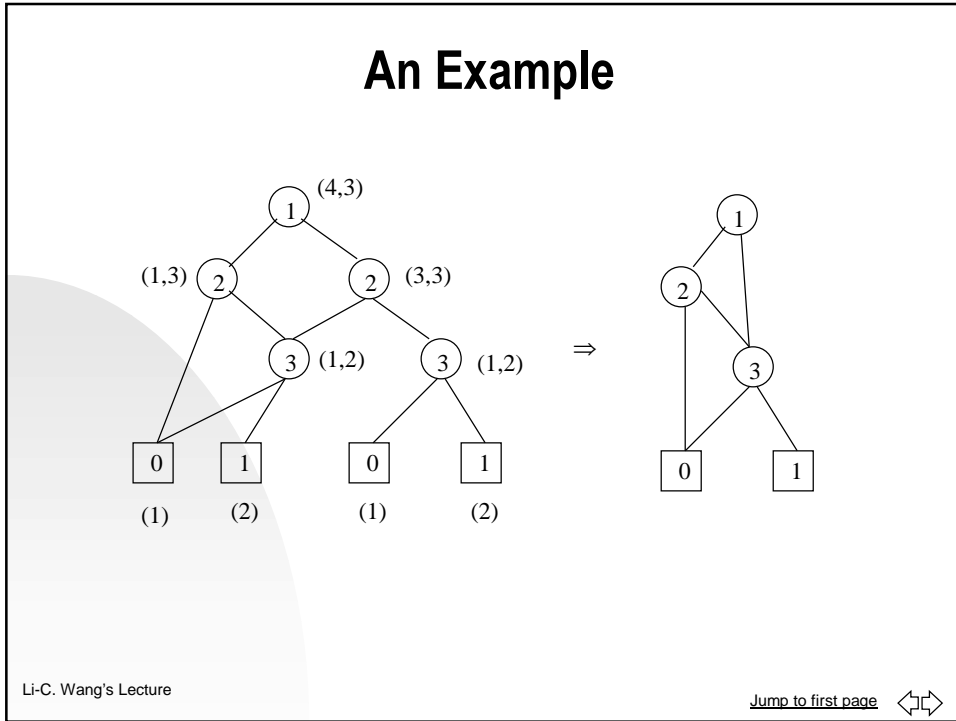
- Why working bottom-up?
 - ◆ So that when working at level i ,
 - ✦ each node below already points to a unique graph already

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Redundant Nodes Removal

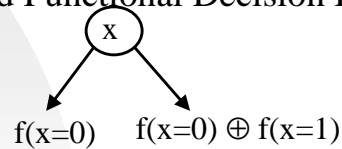
- At each level, how do we know how many nodes are identical in the labeling scheme?
 - ◆ by sorting the labels at that level
 - ◆ $O(n \log n)$ time complexity

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Function Decomposition Rules

- OBDD
 - ◆ $f = x' f(x=0) + x f(x=1)$
- Reed-Muller Decomposition
 - ◆ $f = f(x=0) \oplus x [f(x=0) \oplus f(x=1)]$
 - ◆ $f = f(x=1) \oplus x [f(x=0) \oplus f(x=1)]$
- Ordered Functional Decision Diagram

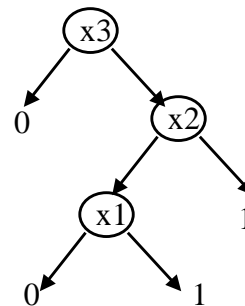


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OFDD

- $F = (x1 \oplus x2) x3$
 - ◆ $F(x3=0) = 0$
 - ◆ $F(x3=1) = x1 \oplus x2$
 - † difference $D3 = x1 \oplus x2$
 - ◆ $D3(x2=0) = x1$
 - ◆ $D3(x2=1) = x1'$
 - † difference $D2 = 1$



- OFDD requires a different Reduction rule
- For some functions, OFDD can be exponentially smaller than OBDD and vice versa
- Evaluation of the graph is different from OBDD
 - ◆ Can no longer be simply tracing from root to 0's or 1's

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Free BDD

- The ordering along each path can be different
- Main issues
 - ◆ Deciding that two trees are representing the same functions can not be done by graph isomorphism.
 - ◆ It can only be done via a probabilistic algorithms
- Free BDD give flexibility in the representation but sacrifice in the complexity of manipulate the BDD data structures

Represent Numeric-Valued Functions

- Arithmetic Decision Diagram (ADD)
 - ◆ Associate more values as terminal nodes
- Edge-Valued BDD (EBDD)
 - ◆ Bring values from the terminal nodes to edges
- Binary Moment Diagram (BMD)
 - ◆ Decomposition : $f = f(x=0) + x (\delta f / \delta x)$

eg. $f = x_0 + 2x_1 + 4x_2$

Linearly Inductive BDD

- Linearly Inductive Function
 - ◆ Recursively Defined
 - ✦ X^i : the inputs to the i -instance function
 - ✦ F^i : the i -instance function
 - ✦ B^i : an arbitrary boolean function
 - ✦ $F^1 = B^1(x^1)$ and $F^i = B^i(X^i, G^{i-1})$
 - ✦ $B^i = B^j$ for all i, j
- Example - Adder
 - ◆ $\text{Sum}^1 = a_1 \oplus b_1 \oplus c_{\text{in}}$
 - ◆ $\text{Sum}^i = a_i \oplus b_i \oplus \text{Carry}^{i-1}$
 - ◆ $\text{Carry}^1 = a_1 b_1 + (a_1 + b_1) c_{\text{in}}$
 - ◆ $\text{Carry}^i = a_i b_i + (a_i + b_i) \text{Carry}^{i-1}$

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LIBDD

- Require different rules for *Apply* and *Reduce*
- Applicable to special case when good sequentiality can be captured
- Function definition has to be known in advance

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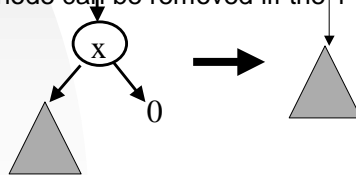
Zero-Suppressed BDD

- Inspiration

- ◆ To represent a collection of bit vectors
 - ✦ eg. (0001) (0100) (1010)
 - ✦ Such a representation is often encountered in some combinatorial problems
- ◆ Usually, the 1's in the bit vectors is sparse

- ZBDD

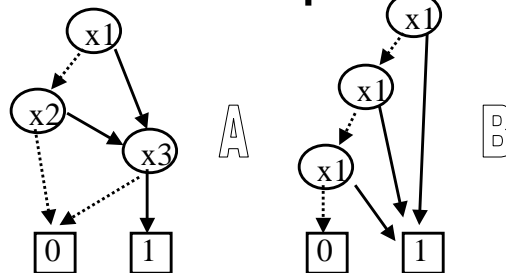
- ◆ Change the Reduction rule
 - ✦ A node can be removed iff the 1-edge points to 0



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ZBDD Interpretation



- In ZBDD sense
 - ◆ $A = \{ 101, 011 \}$ and $B = \{ 100, 010, 001 \}$
- ZBDD is good for sets manipulation
- ZBDD can be used to represent a function as a collection of *cube* sets
 - ◆ $x_1, x_2, x_3 = \{01-, -11, -00\} \Rightarrow \{011000, 001010, 000101\}$
 - ◆ To ensure sparse $\Rightarrow 0: 01, 1: 10, -: 00$

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