

# Introduction to Electronic Design Automation



Jie-Hong Roland Jiang  
江介宏

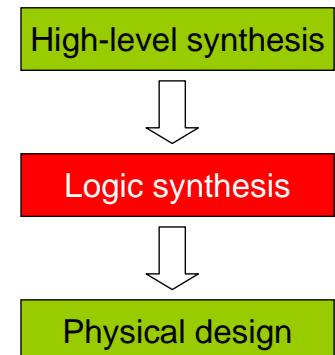
Department of Electrical Engineering  
National Taiwan University



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## Logic Synthesis



Part of the slides are by courtesy of Prof. Andreas Kuehlmann

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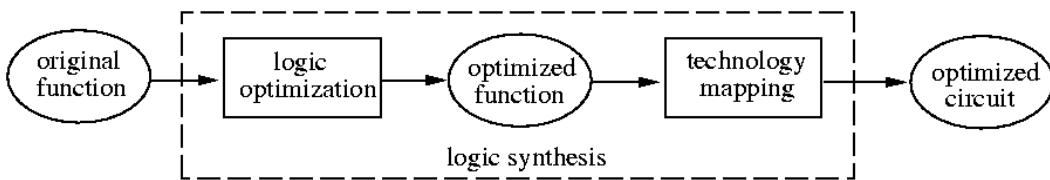
# Logic Synthesis

## □ Course contents

- Overview
- Boolean function representation
- Logic optimization
- Technology mapping

## □ Reading

- Chapter 6



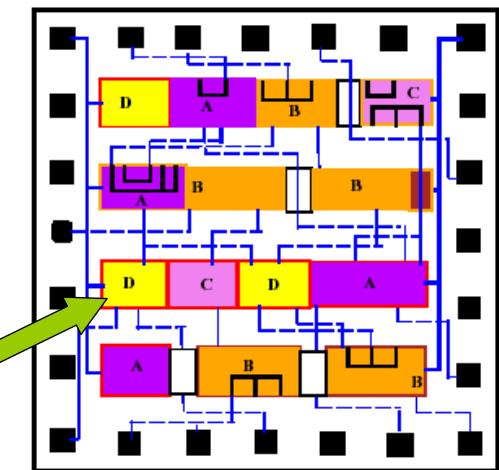
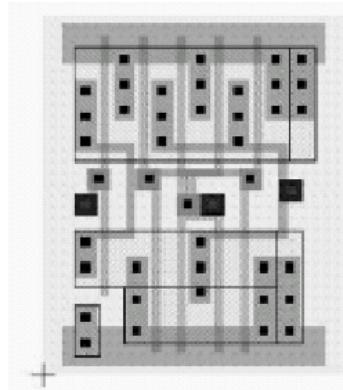
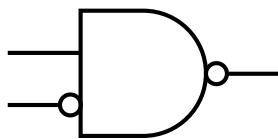
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# High-Level to Logic Synthesis

- Hardware is normally partitioned into two parts:
  - **Data path:** a network of functional units, registers, multiplexers and buses.
  - **Control:** the circuit that takes care of having the data present at the right place at a specific time (i.e. FSM), or of presenting the right instructions to a programmable unit (i.e. microcode).
- High-level synthesis often focuses on data-path optimization
  - The control part is then realized as an FSM
- Logic synthesis often focuses on control-logic optimization
  - Logic synthesis is widely used in application-specific IC (ASIC) design, where standard cell design style is most common

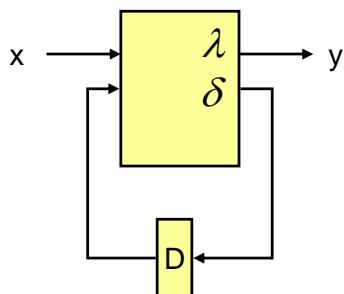
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# Standard-Cell Based Design



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# Transformation of Logic Synthesis



**Given:** Functional description of finite-state machine  $F(Q, X, Y, \delta, \lambda)$  where:

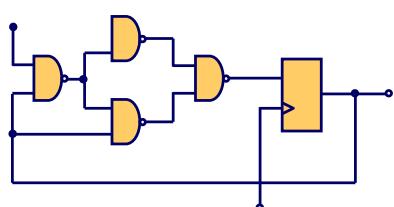
$Q$ : Set of internal states

$X$ : Input alphabet

$Y$ : Output alphabet

$\delta: X \times Q \rightarrow Q$  (next state *function*)

$\lambda: X \times Q \rightarrow Y$  (output *function*)



**Target:** Circuit  $C(G, W)$  where:

$G$ : set of circuit components  $g \in \{\text{gates, FFs, etc.}\}$

$W$ : set of wires connecting  $G$

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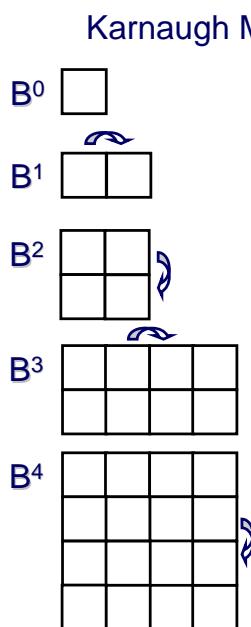
# Boolean Function Representation

- Logic synthesis translates Boolean functions into circuits
- We need representations of Boolean functions for two reasons:
  - to represent and manipulate the actual circuit that we are implementing
  - to facilitate *Boolean reasoning*

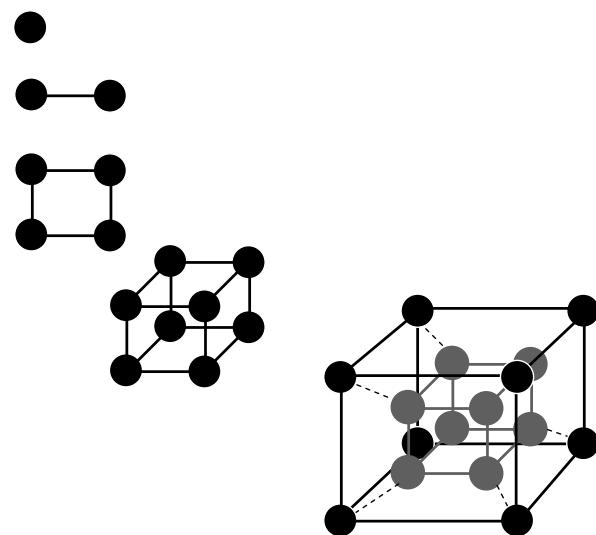
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# Boolean Space

- $B = \{0, 1\}$
- $B^2 = \{0, 1\} \times \{0, 1\} = \{00, 01, 10, 11\}$



Boolean Lattices:



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# Boolean Function

- A Boolean function  $f$  over input variables:  $x_1, x_2, \dots, x_m$  is a mapping  $f: \mathbf{B}^m \rightarrow Y^n$ , where  $\mathbf{B} = \{0,1\}$  and  $Y = \{0,1,d\}$ 
  - E.g.
  - The output value of  $f(x_1, x_2, x_3)$ , say, partitions  $\mathbf{B}^m$  into three sets:
    - **on-set** ( $f=1$ )
      - E.g.  $\{010, 011, 110, 111\}$  (characteristic function  $f^1 = x_2$ )
    - **off-set** ( $f=0$ )
      - E.g.  $\{100, 101\}$  (characteristic function  $f^0 = x_1 \neg x_2$ )
    - **don't-care set** ( $f=d$ )
      - E.g.  $\{000, 001\}$  (characteristic function  $f^d = \neg x_1 \neg x_2$ )
- $f$  is an **incompletely specified function** if the don't-care set is nonempty. Otherwise,  $f$  is a **completely specified function**
  - Unless otherwise said, a Boolean function is meant to be completely specified

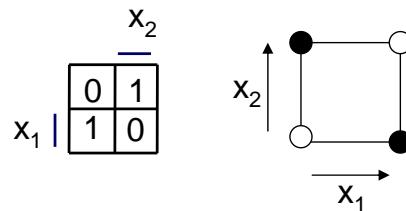
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# Boolean Function

- A Boolean function  $f: \mathbf{B}^n \rightarrow \mathbf{B}$  over variables  $x_1, \dots, x_n$  maps each Boolean valuation (truth assignment) in  $\mathbf{B}^n$  to 0 or 1

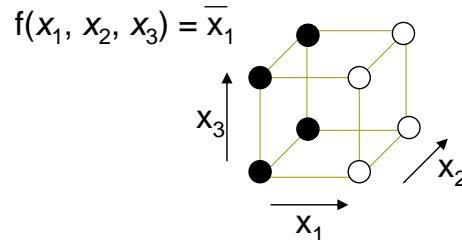
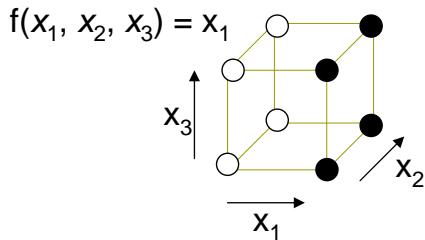
## Example

$f(x_1, x_2)$  with  $f(0,0) = 0, f(0,1) = 1, f(1,0) = 1, f(1,1) = 0$



# Boolean Function

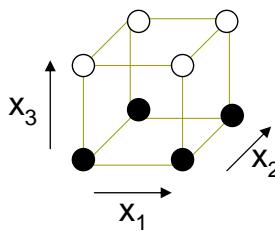
- **Onset** of  $f$ , denoted as  $f^1$ , is  $f^1 = \{v \in \mathbf{B}^n \mid f(v) = 1\}$ 
  - If  $f^1 = \mathbf{B}^n$ ,  $f$  is a **tautology**
- **Offset** of  $f$ , denoted as  $f^0$ , is  $f^0 = \{v \in \mathbf{B}^n \mid f(v) = 0\}$ 
  - If  $f^0 = \mathbf{B}^n$ ,  $f$  is **unsatisfiable**. Otherwise,  $f$  is **satisfiable**.
- $f^1$  and  $f^0$  are sets, not functions!
- Boolean functions  $f$  and  $g$  are **equivalent** if  $\forall v \in \mathbf{B}^n. f(v) = g(v)$  where  $v$  is a truth assignment or Boolean valuation
- A **literal** is a Boolean variable  $x$  or its negation  $x'$  (or  $x, \neg x$ ) in a Boolean formula



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# Boolean Function

- There are  $2^n$  vertices in  $\mathbf{B}^n$
- There are  $2^{2^n}$  distinct Boolean functions
  - Each subset  $f^1 \subseteq \mathbf{B}^n$  of vertices in  $\mathbf{B}^n$  forms a distinct Boolean function  $f$  with onset  $f^1$



$x_1 x_2 x_3$	$f$
0 0 0	1
0 0 1	0
0 1 0	1
0 1 1	0
1 0 0	$\Rightarrow 1$
1 0 1	0
1 1 0	1
1 1 1	0

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# Boolean Operations

Given two Boolean functions:

$$f : \mathbf{B}^n \rightarrow \mathbf{B}$$

$$g : \mathbf{B}^n \rightarrow \mathbf{B}$$

- $h = f \wedge g$  from **AND** operation is defined as

$$h^1 = f^1 \cap g^1; h^0 = \mathbf{B}^n \setminus h^1$$

- $h = f \vee g$  from **OR** operation is defined as

$$h^1 = f^1 \cup g^1; h^0 = \mathbf{B}^n \setminus h^1$$

- $h = \neg f$  from **COMPLEMENT** operation is defined as

$$h^1 = f^0; h^0 = f^1$$

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# Cofactor and Quantification

Given a Boolean function:

$f : \mathbf{B}^n \rightarrow \mathbf{B}$ , with the input variable  $(x_1, x_2, \dots, x_i, \dots, x_n)$

- **Positive cofactor on variable  $x_i$**

$h = f_{x_i}$  is defined as  $h = f(x_1, x_2, \dots, 1, \dots, x_n)$

- **Negative cofactor on variable  $x_i$**

$h = f_{\neg x_i}$  is defined as  $h = f(x_1, x_2, \dots, 0, \dots, x_n)$

- **Existential quantification over variable  $x_i$**

$h = \exists x_i. f$  is defined as  $h = f(x_1, x_2, \dots, 0, \dots, x_n) \vee f(x_1, x_2, \dots, 1, \dots, x_n)$

- **Universal quantification over variable  $x_i$**

$h = \forall x_i. f$  is defined as  $h = f(x_1, x_2, \dots, 0, \dots, x_n) \wedge f(x_1, x_2, \dots, 1, \dots, x_n)$

- **Boolean difference over variable  $x_i$**

$h = \partial f / \partial x_i$  is defined as  $h = f(x_1, x_2, \dots, 0, \dots, x_n) \oplus f(x_1, x_2, \dots, 1, \dots, x_n)$

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# Boolean Function Representation

- Some common representations:
  - Truth table
  - Boolean formula
    - SOP (sum-of-products, or called disjunctive normal form, DNF)
    - POS (product-of-sums, or called conjunctive normal form, CNF)
  - BDD (binary decision diagram)
  - Boolean network (consists of nodes and wires)
    - Generic Boolean network
      - Network of nodes with generic functional representations or even subcircuits
    - Specialized Boolean network
      - Network of nodes with SOPs (PLAs)
      - And-Inv Graph (AIG)
- Why different representations?
  - Different representations have their own strengths and weaknesses (no single data structure is best for all applications)

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## Boolean Function Representation Truth Table

- Truth table (function table for multi-valued functions):

The **truth table** of a function  $f : \mathbf{B}^n \rightarrow \mathbf{B}$  is a tabulation of its value at each of the  $2^n$  vertices of  $\mathbf{B}^n$ .

In other words the truth table lists all **mintems**

Example:  $f = a'b'c'd + a'b'cd + a'bc'd + ab'c'd + ab'cd + abc'd + abcd' + abcd$

The truth table representation is

- impractical for large  $n$
- canonical

If two functions are the equal, then their **canonical** representations are isomorphic.

	abcd	f		abcd	f
0	0000	0	8	1000	0
1	0001	1	9	1001	1
2	0010	0	10	1010	0
3	0011	1	11	1011	1
4	0100	0	12	1100	0
5	0101	1	13	1101	1
6	0110	0	14	1110	1
7	0111	0	15	1111	1

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# Boolean Function Representation

## Boolean Formula

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- A **Boolean formula** is defined inductively as an expression with the following formation rules (syntax):

formula ::=	'(' formula ')	
	Boolean constant	(true or false)
	<Boolean variable>	
	formula "+" formula	(OR operator)
	formula "." formula	(AND operator)
	$\neg$ formula	(complement)

### Example

$$f = (x_1 \cdot x_2) + (x_3) + \neg(\neg(x_4 \cdot \neg x_1))$$

typically “.” is omitted and ‘(, )’ are omitted when the operator priority is clear, e.g.,  $f = x_1 x_2 + x_3 + x_4 \neg x_1$

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# Boolean Function Representation

## Boolean Formula in SOP

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- Any function can be represented as a **sum-of-products (SOP)**, also called **sum-of-cubes** (a **cube** is a product term), or **disjunctive normal form (DNF)**

### Example

$$\varphi = ab + a'c + bc$$

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# Boolean Function Representation Boolean Formula in POS

- Any function can be represented as a **product-of-sums (POS)**, also called **conjunctive normal form (CNF)**
  - Dual of the SOP representation

## Example

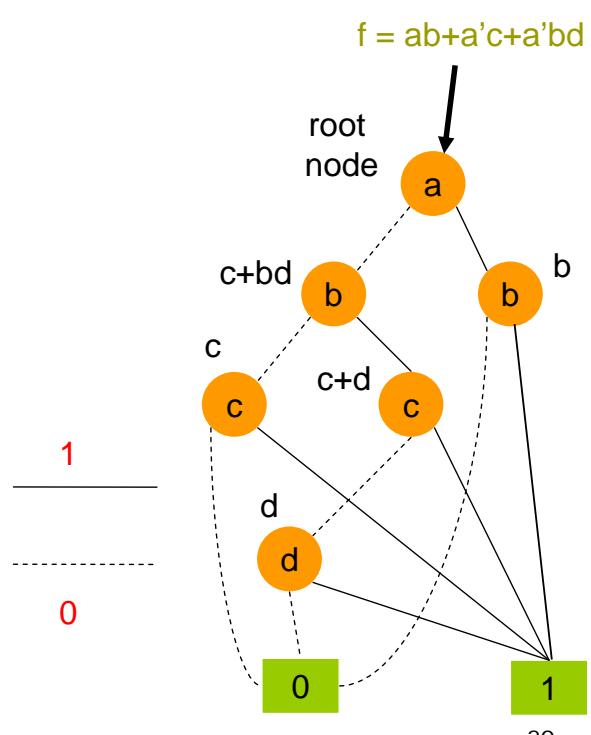
$$\varphi = (a+b'+c) (a'+b+c) (a+b'+c') (a+b+c)$$

- Exercise: Any Boolean function in POS can be converted to SOP using De Morgan's law and the distributive law, and vice versa

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# Boolean Function Representation Binary Decision Diagram

- BDD – a graph representation of Boolean functions
  - A **leaf node** represents constant 0 or 1
  - A **non-leaf node** represents a decision node (multiplexer) controlled by some variable
  - Can make a BDD representation **canonical** by imposing the **variable ordering** and **reduction** criteria (ROBDD)



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# Boolean Function Representation

## Binary Decision Diagram

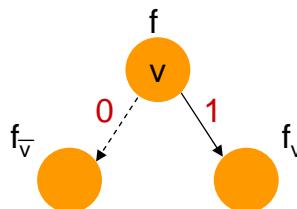
- Any Boolean function  $f$  can be written in term of **Shannon expansion**

$$f = v f_v + \neg v f_{\neg v}$$

- Positive cofactor:  $f_{x_i} = f(x_1, \dots, x_i=1, \dots, x_n)$
- Negative cofactor:  $f_{\neg x_i} = f(x_1, \dots, x_i=0, \dots, x_n)$

- BDD is a compressed Shannon cofactor tree:

- The two children of a node with function  $f$  controlled by variable  $v$  represent two sub-functions  $f_v$  and  $f_{\neg v}$



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# Boolean Function Representation

## Binary Decision Diagram

- Reduced and ordered BDD (ROBDD) is a **canonical** Boolean function representation

- Ordered:

- cofactor variables are in the **same order along all paths**

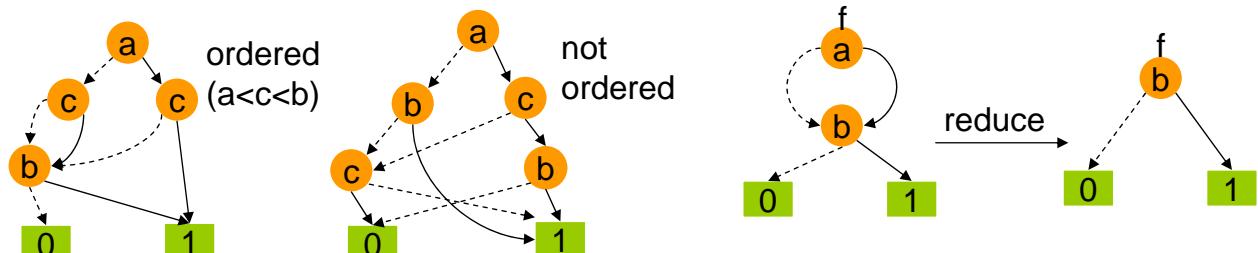
$$x_{i_1} < x_{i_2} < x_{i_3} < \dots < x_{i_n}$$

- Reduced:

- any node with two identical children is removed

- two nodes with isomorphic BDD's are merged

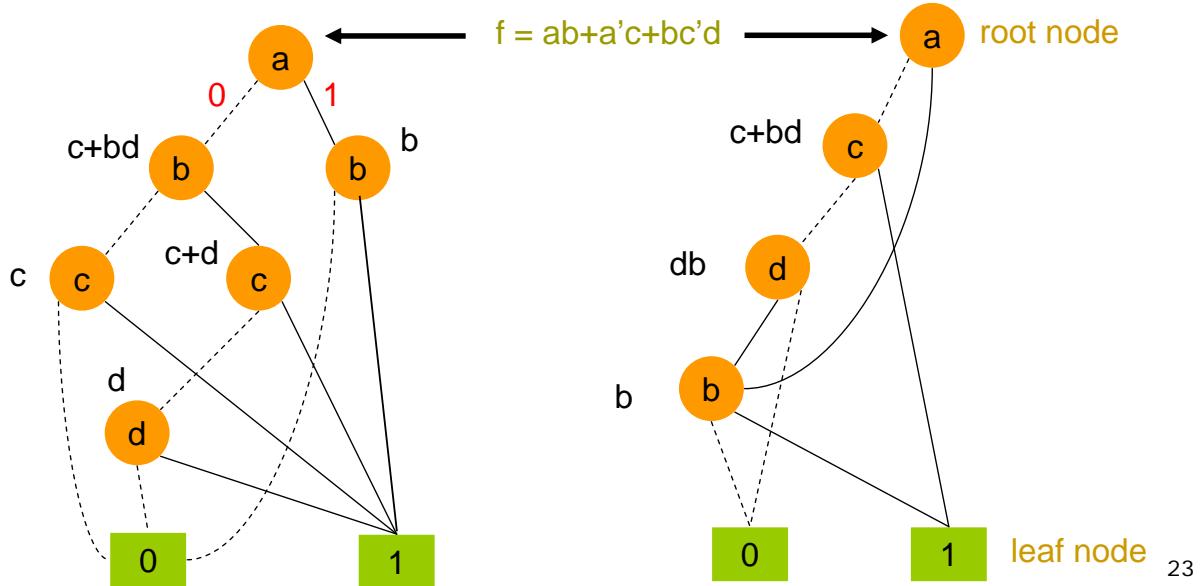
These two rules make any node in an ROBDD represent a distinct logic function



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# Boolean Function Representation Binary Decision Diagram

- For a Boolean function,
  - ROBDD is unique with respect to a given variable ordering
  - Different orderings may result in different ROBDD structures



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# Boolean Function Representation Boolean Network

- A **Boolean network** is a directed graph  $C(G, N)$  where  $G$  are the gates and  $N \subseteq (G \times G)$  are the directed edges (nets) connecting the gates.

Some of the vertices are designated:

**Inputs:**  $I \subseteq G$

**Outputs:**  $O \subseteq G$

$$I \cap O = \emptyset$$

Each gate  $g$  is assigned a Boolean function  $f_g$  which computes the output of the gate in terms of its inputs.

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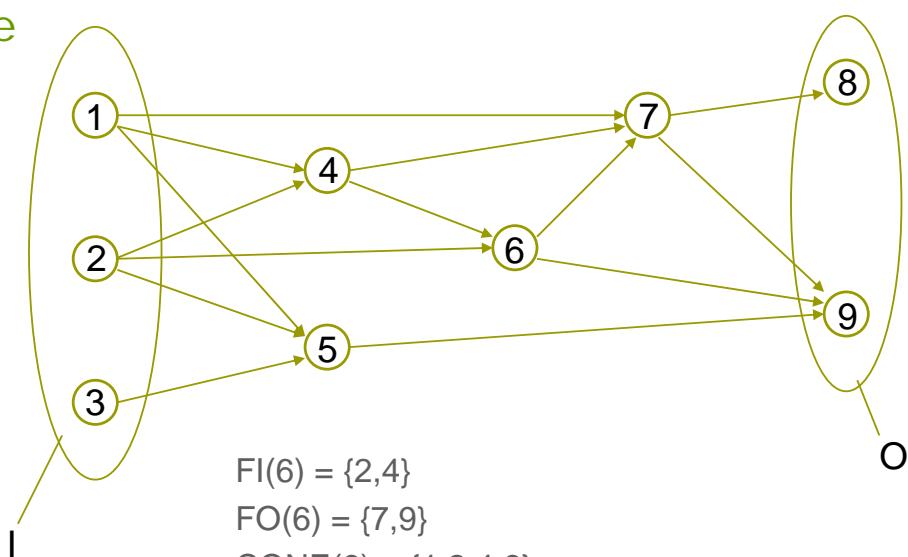
# Boolean Function Representation Boolean Network

- The **fanin**  $FI(g)$  of a gate  $g$  are the predecessor gates of  $g$ :  
 $FI(g) = \{g' \mid (g',g) \in N\}$  ( $N$ : the set of nets)
- The **fanout**  $FO(g)$  of a gate  $g$  are the successor gates of  $g$ :  
 $FO(g) = \{g' \mid (g,g') \in N\}$
- The **cone**  $CONE(g)$  of a gate  $g$  is the **transitive fanin** (TFI) of  $g$  and  $g$  itself
- The **support**  $SUPPORT(g)$  of a gate  $g$  are all inputs in its cone:  
 $SUPPORT(g) = CONE(g) \cap I$

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# Boolean Function Representation Boolean Network

## Example

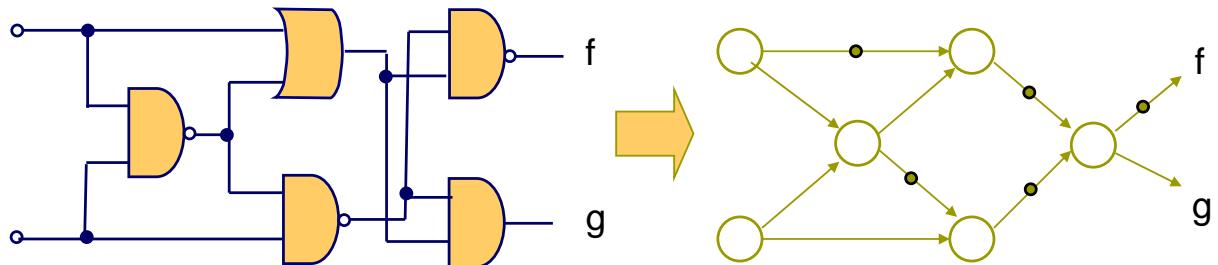


Every node may have its own function

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# Boolean Function Representation And-Inverter Graph

- AND-INVERTER graphs (AIGs)
  - vertices: 2-input AND gates
  - edges: interconnects with (optional) dots representing INVs
- Hash table to identify and reuse structurally isomorphic circuits



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# Boolean Function Representation

- A **canonical form** of a Boolean function is a **unique** representation of the function
  - It can be used for verification purposes
- Example**
  - Truth table is canonical
    - It grows exponentially with the number of input variables
  - ROBDD is canonical
    - It is of practical interests because it may represent many Boolean functions compactly
  - SOP, POS, Boolean networks are NOT canonical

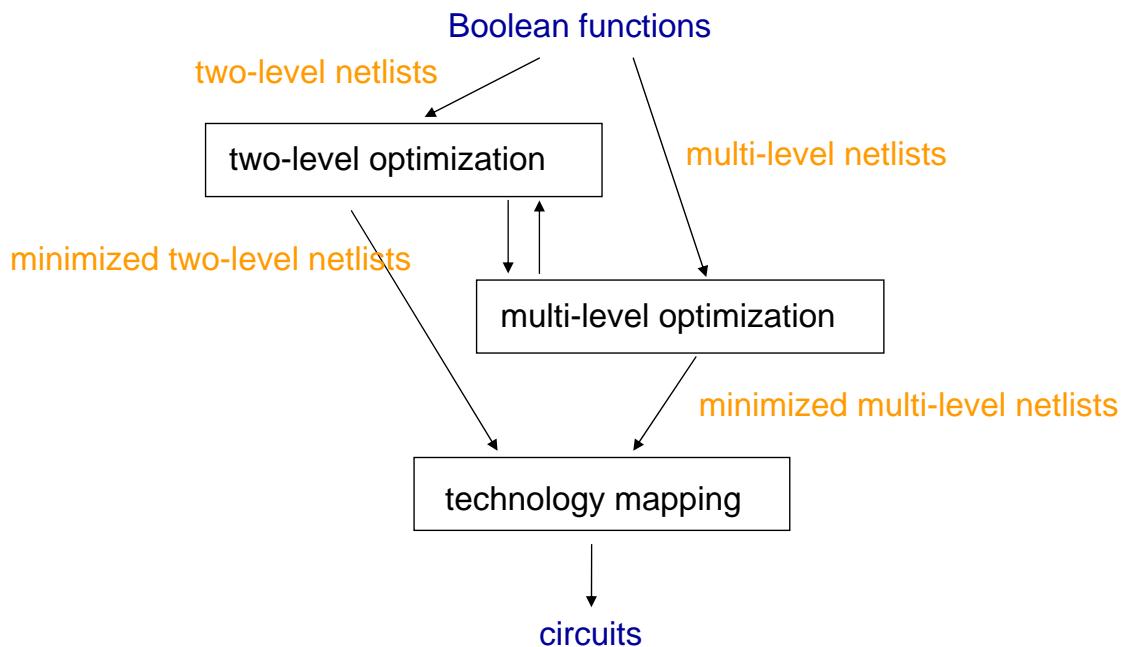
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# Boolean Function Representation

- ❑ Truth table
  - Canonical
  - Useful in representing small functions
- ❑ SOP
  - Useful in two-level logic optimization, and in representing local node functions in a Boolean network
- ❑ POS
  - Useful in SAT solving and Boolean reasoning
  - Rarely used in circuit synthesis (due to the asymmetric characteristics of NMOS and PMOS)
- ❑ ROBDD
  - Canonical
  - Useful in Boolean reasoning
- ❑ Boolean network
  - Useful in multi-level logic optimization
- ❑ AIG
  - Useful in multi-level logic optimization and Boolean reasoning

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# Logic Optimization



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# Two-Level Logic Minimization

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- Any Boolean function can be realized using PLA in two levels: AND-OR (sum of products), NAND-NAND, etc.
  - Direct implementation of two-level logic using PLAs (programmable logic arrays) is not as popular as in the nMOS days
- Classic problem solved by the *Quine-McCluskey* algorithm
  - Popular cost function: #cubes and #literals in an SOP expression
    - #cubes – #rows in a PLA
    - #literals – #transistors in a PLA
  - The goal is to find a **minimal irredundant prime cover**

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# Two-Level Logic Minimization

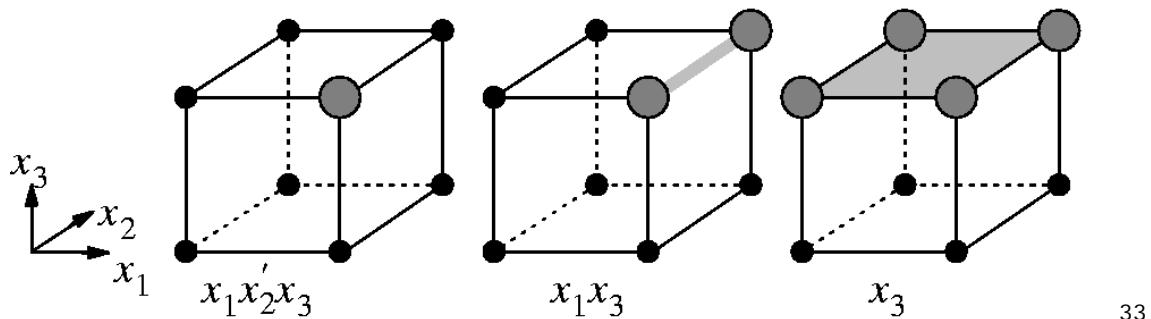
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- Exact algorithm
  - Quine-McCluskey's procedure
- Heuristic algorithm
  - Espresso

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# Two-Level Logic Minimization Minterms and Cubes

- A **minterm** is a product of **every** input variable or its negation
  - A minterm corresponds to a single point in  $\mathbf{B}^n$
- A **cube** is a product of literals
  - The fewer the number of literals is in the product, the bigger the space is covered by the cube



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# Two-Level Logic Minimization Implicant and Cover

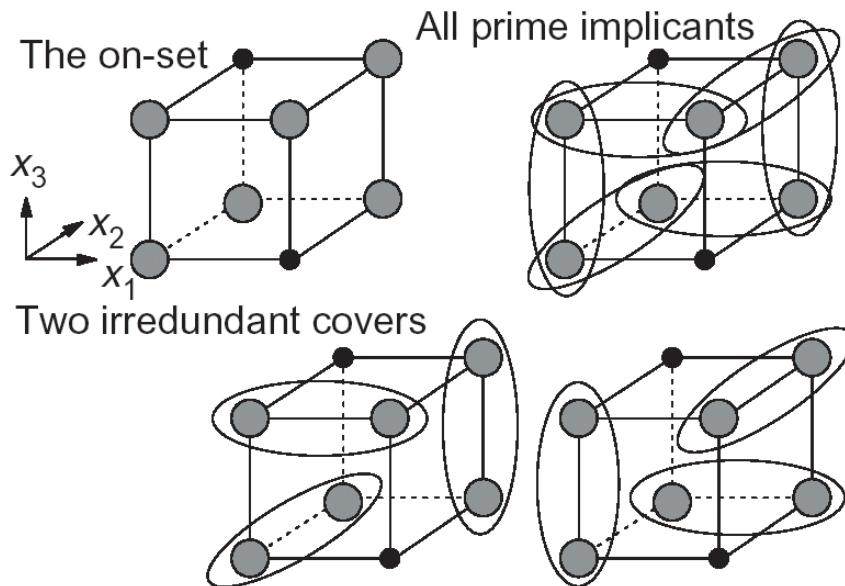
- An **implicant** is a cube whose points are either in the on-set or the dc-set.
- A **prime implicant** is an implicant that is not included in any other implicant.
- A set of prime implicants that together cover all points in the on-set (and some or all points of the dc-set) is called a **prime cover**.
  - A prime cover is **irredundant** when none of its prime implicants can be removed from the cover.
  - An irredundant prime cover is **minimal** when the cover has the minimal number of prime implicants.  
(c.f. minimum vs. minimal)

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# Two-Level Logic Minimization Cover

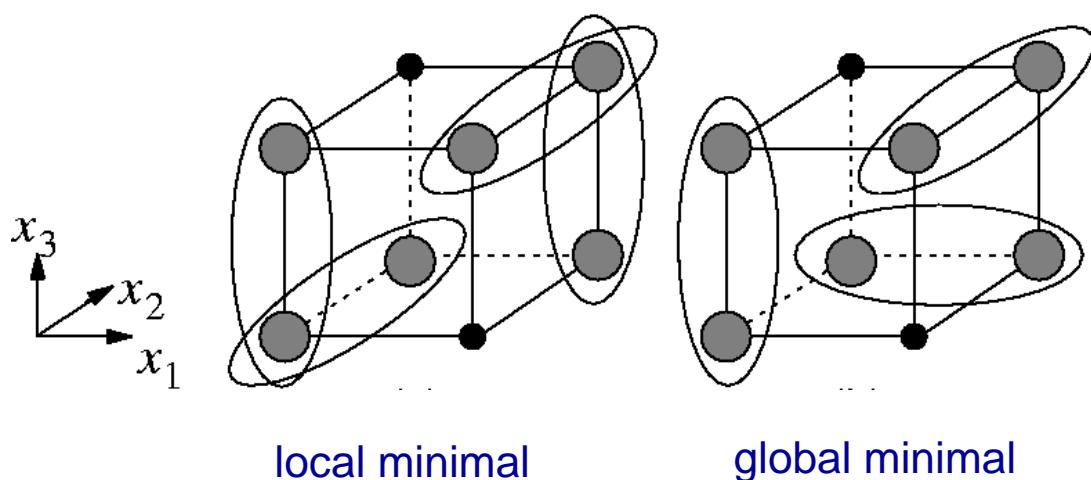
## Example

- $f = \neg x_1 \neg x_3 + \neg x_2 x_3 + x_1 x_2$
- $f = \neg x_1 \neg x_2 + x_2 \neg x_3 + x_1 x_3$



# Two-Level Logic Minimization Cover

## Example



# Two-Level Logic Minimization

## Quine-McCluskey Procedure

- Given  $G$  and  $D$  (covers for  $\mathfrak{I} = (f, d, r)$  and  $d$ , respectively), find a minimum cover  $G^*$  of primes where:
 
$$f \subseteq G^* \subseteq f+d \quad (G^* \text{ is a prime cover of } \mathfrak{I})$$
  - $f$  is the onset,  $d$  don't-care set, and  $r$  offset
- Q-M Procedure:
  - Generate all primes of  $\mathfrak{I}$ ,  $\{P_j\}$  (i.e. primes of  $(f+d) = G+D$ )
  - Generate all minterms  $\{m_i\}$  of  $f = G \wedge \neg D$
  - Build Boolean matrix  $B$  where
 
$$B_{ij} = 1 \text{ if } m_i \in P_j$$

$$= 0 \text{ otherwise}$$
  - Solve the minimum column covering problem for  $B$  (unate covering problem)

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# Two-Level Logic Minimization

## Quine-McCluskey Procedure

### Generating Primes

Tabular method  
(based on *consensus* operation):

- Start with all **minterm canonical form** of  $F$
- Group *pairs* of adjacent minterms into cubes
- Repeat merging cubes until no more merging possible; mark ( $\checkmark$ ) + remove all covered cubes.
- Result: set of *primes* of  $f$ .

### Example

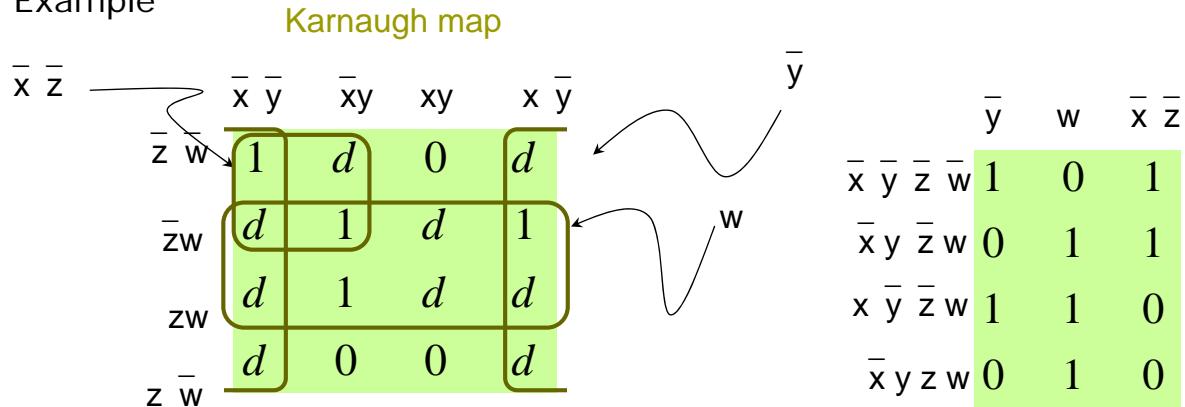
$$F = x'y' + wx'y + x'y'z' + w'y'z$$

$$F = x'y' + wx'y + x'y'z' + w'y'z$$

$w'x'y'z'$	$\checkmark$	$w'x'y'$	$\checkmark$	$x'y'$
		$w'x'z'$	$\checkmark$	$x'z'$
		$x'y'z'$	$\checkmark$	
$w'x'y'z$	$\checkmark$	$x'y'z$	$\checkmark$	
$w'x'y'z'$	$\checkmark$	$x'y'z'$	$\checkmark$	
$wx'y'z'$	$\checkmark$	$wx'y'$	$\checkmark$	
		$wx'z'$	$\checkmark$	
$wx'y'z$	$\checkmark$	$wy'z$		
$wx'y'z'$	$\checkmark$	$wyz'$		
$wxyz'$	$\checkmark$	$wxy$		
$wxy'z$	$\checkmark$	$wxz$		
$wxyz$	$\checkmark$			

# Two-Level Logic Minimization Quine-McCluskey Procedure

## Example



$$F = \bar{x} y z w + \bar{x} y \bar{z} w + x \bar{y} \bar{z} w + \bar{x} y z w \quad (\text{cover of } \mathfrak{J})$$

$$D = \bar{y} z + x y w + \bar{x} y z w + x \bar{y} w + \bar{x} y z w \quad (\text{cover of } d)$$

Primes:  $\bar{y} + w + \bar{x} \bar{z}$

## Covering Table

Solution:  $\{1, 2\} \Rightarrow \bar{y} + w$  is a minimum prime cover (also  $w + \bar{x} \bar{z}$ )

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# Two-Level Logic Minimization Quine-McCluskey Procedure

## Column covering of Boolean matrix

	$\bar{y}$	w	$\bar{x} \bar{z}$	Primes of $f+d$
$\bar{x} \bar{y} \bar{z} \bar{w}$	1	0	1	
$\bar{x} y \bar{z} w$	0	1	1	
$x \bar{y} \bar{z} w$	1	1	0	
$\bar{x} y z w$	0	1	0	Row singleton (essential minterm)

Essential prime

↑

Minterms of  $f$

- Definition. An **essential prime** is a prime that covers an onset minterm of  $f$  not covered by any other primes.

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# Two-Level Logic Minimization

## Quine-McCluskey Procedure

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### □ Row equality in Boolean matrix:

- In practice, many rows in a covering table are identical. That is, there exist minterms that are contained in the same set of primes.

#### ■ Example

$m_1$	0101101
$m_2$	0101101

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# Two-Level Logic Minimization

## Quine-McCluskey Procedure

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### □ Row dominance in Boolean matrix:

- A *row*  $i_1$  whose set of primes is contained in the set of primes of *row*  $i_2$  is said to **dominate**  $i_2$ .

#### ■ Example

$i_1$	011010
$i_2$	011110

□  $i_1$  dominates  $i_2$

□ Can remove row  $i_2$  because have to choose a prime to cover  $i_1$ , and any such prime also covers  $i_2$ . So  $i_2$  is automatically covered.

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# Two-Level Logic Minimization

## Quine-McCluskey Procedure

- Column dominance in Boolean matrix:
  - A *column*  $j_1$  whose rows are a superset of another *column*  $j_2$  is said to **dominate**  $j_2$ .
  - Example
- |   | $j_1$ | $j_2$ |
|---|-------|-------|
| 1 | 0     |       |
| 0 | 0     |       |
| 1 | 1     |       |
| 0 | 0     |       |
| 1 | 1     |       |
- $j_1$  dominates  $j_2$
- We can remove column  $j_2$  since  $j_1$  covers all those rows and more. We would never choose  $j_2$  in a minimum cover since it can always be replaced by  $j_1$ .

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# Two-Level Logic Minimization

## Quine-McCluskey Procedure

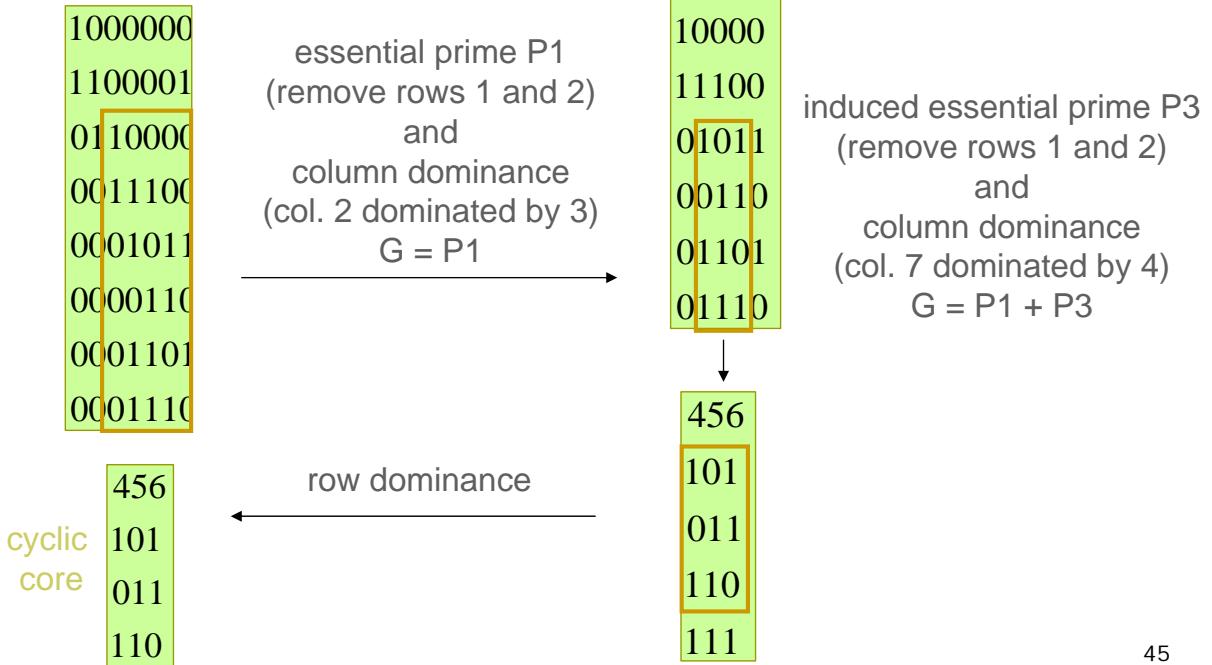
- Reducing Boolean matrix
  - 1. Remove all rows covered by essential primes (columns in row singletons). Put these primes in the cover  $G$ .
  - 2. Group identical rows together and remove dominated rows.
  - 3. Remove dominated columns. For equal columns, keep one prime to represent them.
  - 4. Newly formed row singletons define **induced essential primes**.
  - 5. Go to 1 if covering table decreased.
- The resulting reduced covering table is called the **cyclic core**. This has to be solved (**unate covering problem**). A minimum solution is added to  $G$ . The resulting  $G$  is a minimum cover.

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# Two-Level Logic Minimization

## Quine-McCluskey Procedure

Example (reducing Boolean matrix)



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# Two-Level Logic Minimization

## Quine-McCluskey Procedure

Solving cyclic core

- ❑ Best known method (for unate covering) is **branch and bound** with some clever bounding heuristics
- ❑ Independent Set Heuristic:
  - Find a maximum set  $I$  of “independent” rows. Two rows  $B_{i_1}, B_{i_2}$  are independent if **not**  $\exists j$  such that  $B_{i_1j} = B_{i_2j} = 1$ . (They have no column in common.)

Example

A covering matrix B rearranged with independent sets first

$$B = \begin{array}{|c|c|c|c|} \hline & 1 & 1 & 1 \\ & & 1 & 1 & 1 & 1 \\ & & & 1 & 1 & & 0 \\ \hline & & & & & & \\ \hline & A & & & & C & \\ \hline \end{array} \quad \left. \right\} \text{Independent set } \mathcal{I} \text{ of rows}$$

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# Two-Level Logic Minimization Quine-McCluskey Procedure

## Solving cyclic core

### □ Heuristic algorithm:

- Let  $\mathcal{J} = \{I_1, I_2, \dots, I_k\}$  be the independent set of rows
1. choose  $j \in \mathcal{J}$  such that column  $j$  covers the most rows of  $A$ . Put  $P_j$  in  $G$
  2. eliminate all rows covered by column  $j$
  3.  $\mathcal{J} \leftarrow \mathcal{J} \setminus \{I_j\}$
  4. go to 1 if  $|\mathcal{J}| > 0$
  5. If  $B$  is empty, then done (in this case achieve minimum solution)
  6. If  $B$  is not empty, choose an independent set of  $B$  and go to 1

1 1 1	1 1 1 1	1 1	0
A			C

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# Two-Level Logic Minimization Quine-McCluskey Procedure

## □ Summary

- Calculate all prime implicants (of the union of the onset and don't care set)
- Find the minimal cover of all minterms in the onset by prime implicants
  - Construct the covering matrix
  - Simplify the covering matrix by detecting **essential** columns, **row and column dominance**
  - What is left is the **cyclic core** of the covering matrix.
    - The covering problem can then be solved by a branch-and-bound algorithm.

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# Two-Level Logic Minimization Exact vs. Heuristic Algorithms

## □ Quine-McCluskey Method:

1. Generate cover of all primes  $G = p_1 + p_2 + \dots + p_{3^n/n}$
2. Make G irredundant (in optimum way)
  - Q-M is **exact**, i.e., it gives an exact minimum

## □ Heuristic Methods:

1. Generate (somehow) a cover of  $\mathfrak{I}$  using some of the primes  $G = p_{i_1} + p_{i_2} + \dots + p_{i_k}$
2. Make G irredundant (maybe not optimally)
3. Keep best result - try again (i.e. go to 1)

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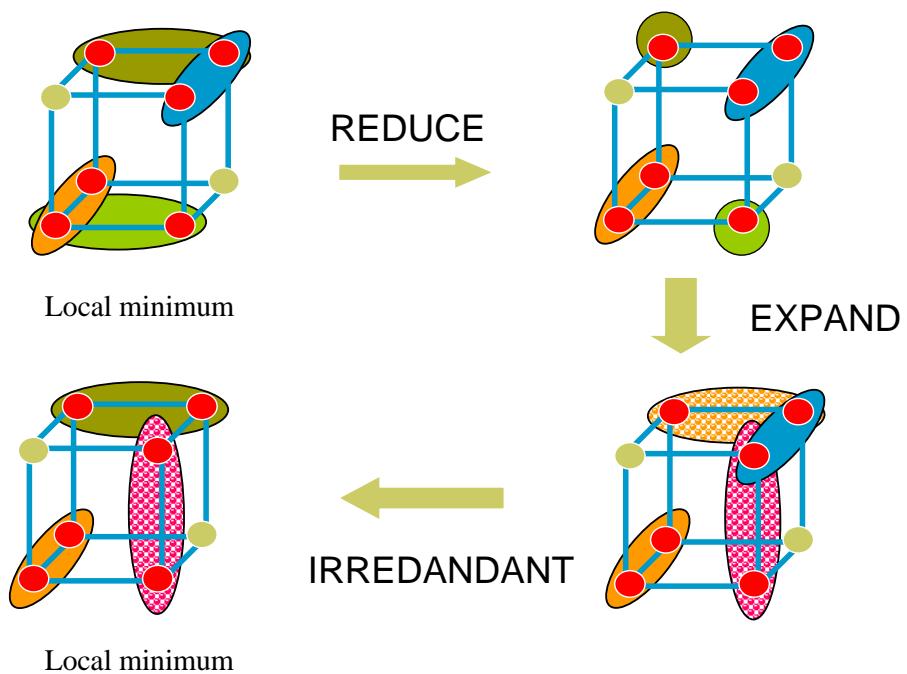
# Two-Level Logic Minimization ESPRESSO

## □ Heuristic two-level logic minimization ESPRESSO( $\mathfrak{I}$ )

```
{  
  (F,D,R) ← DECODE( $\mathfrak{I}$ )  
  F ← EXPAND(F,R)                                //LASTGASP  
  F ← IRREDUNDANT(F,D)  
  E ← ESSENTIAL_PRIMES(F,D)  
  F ← F-E; D ← D+E  
  do{  
    do{  
      F ← REDUCE(F,D)  
      F ← EXPAND(F,R)  
      F ← IRREDUNDANT(F,D)  
    } while fewer terms in F  
    F ← F+E; D ← D-E  
    LOWER_OUTPUT(F,D)  
    RAISE_INPUTS(F,R)  
    error ← ( $F_{\text{old}} \not\subset F$ ) or ( $F \not\subset F_{\text{old}} + D$ )  
  } return (F,error)  
}
```

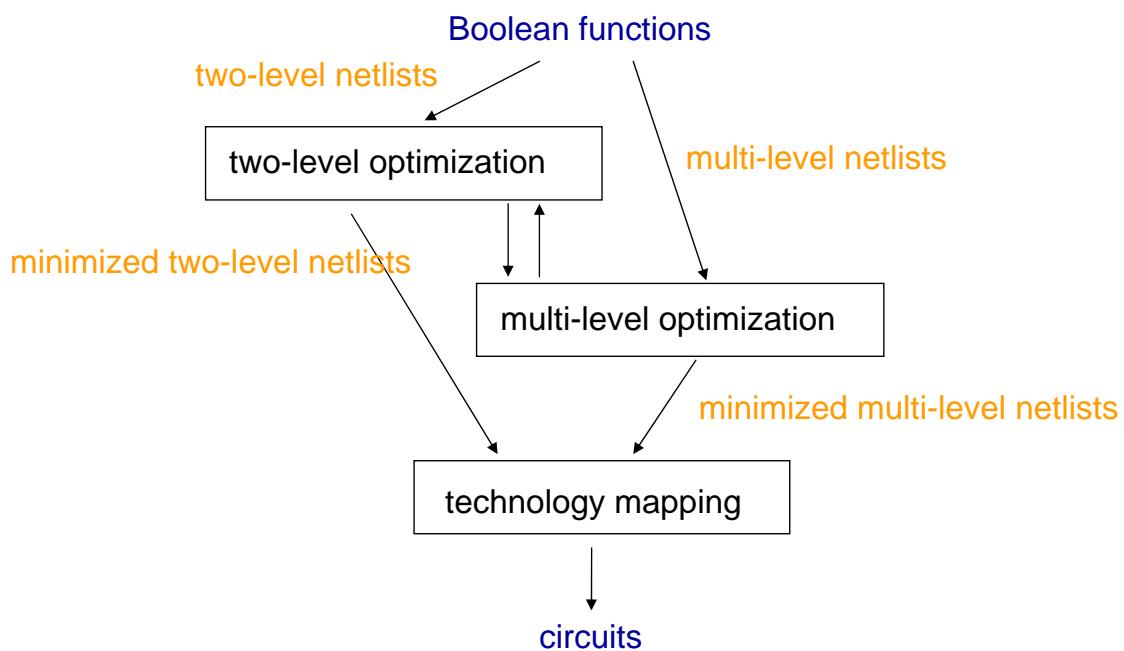
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# Two-Level Logic Minimization ESPRESSO



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# Logic Minimization



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