

Two-Level Logic Minimization

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Topics

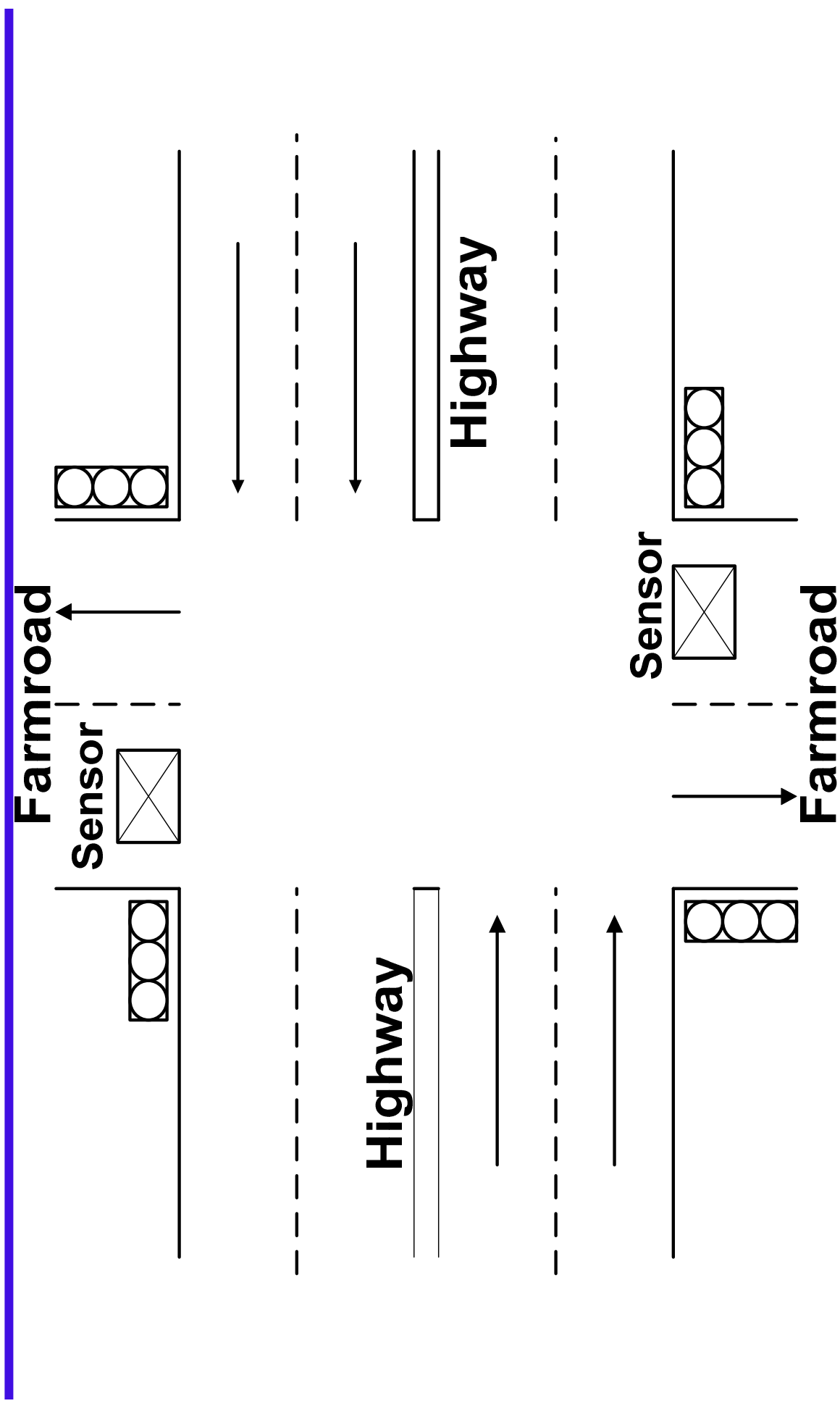
- **Motivation**
- **Boolean functions & notation**
- **Exact 2-level logic minimization**
 - **Quine-McCluskey**
- **Heuristic 2-level minimization**
 - **MINI, Espresso**

Schematic Entry Era

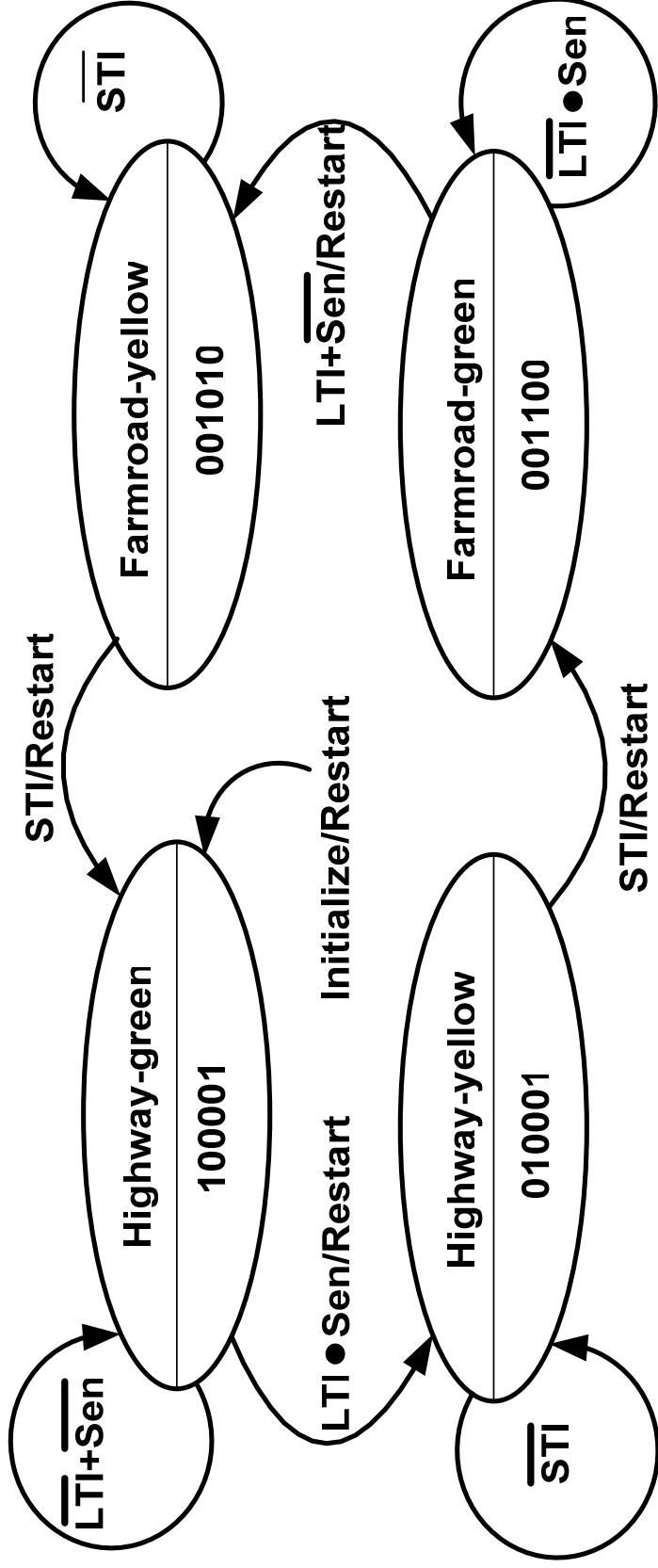
Given:

- Gate-level schematic entry editor
- Gate-level simulator (we haven't talked about this)
- Gate level static-timing analyzer
- Netlist → Layout flow
- We can (and did) build large-scale integrated (35,000 gate) circuits
- EDA vendors provided front-end tools and ASIC vendor (e.g. LSI Logic) provided back-end flow
- But ... It may be much more natural, and productive, to describe complex control logic by Boolean equations than by a schematic netlist of gates

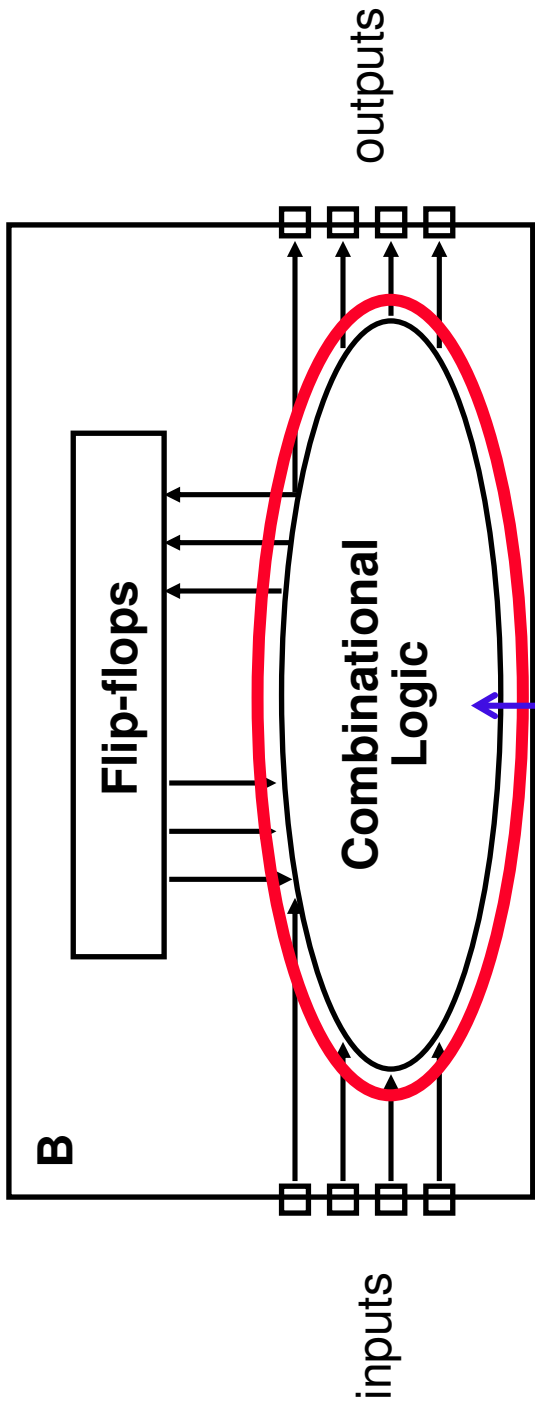
For example: traffic light controller



As a State transition diagram

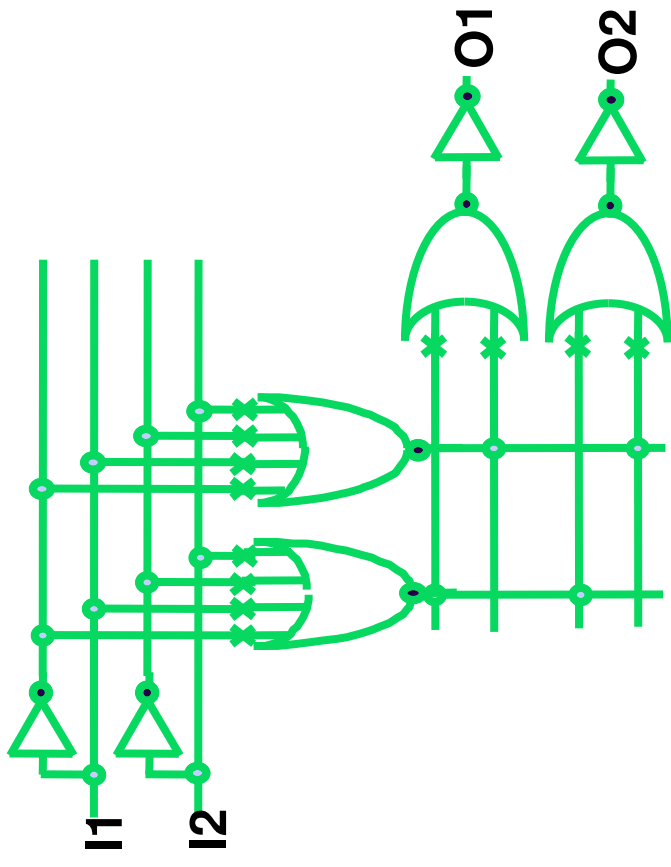
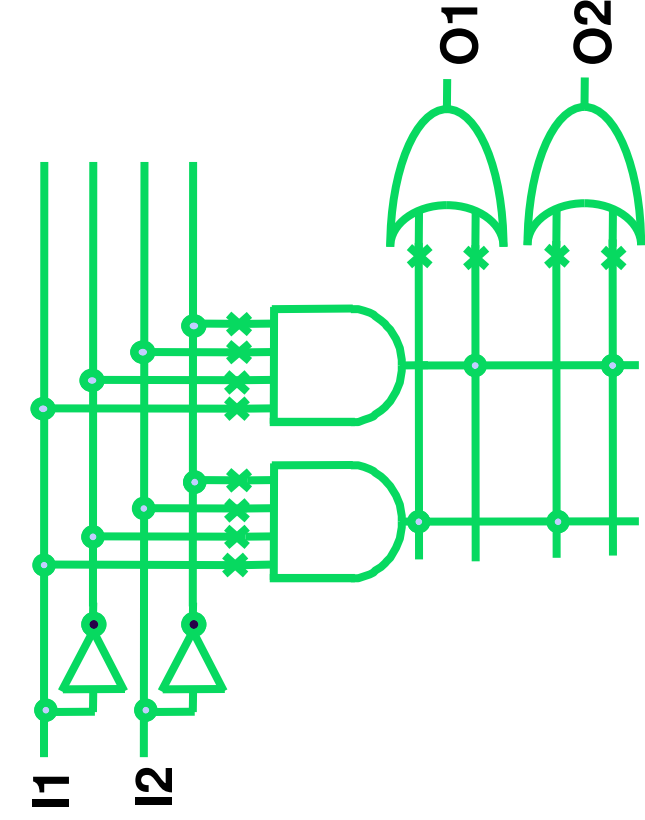


Synthesize Logic to Implement equations



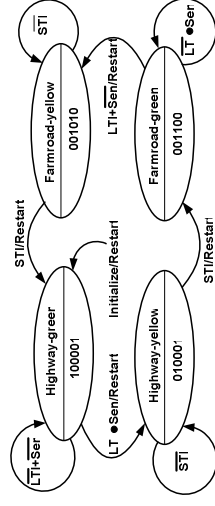
Describe using Boolean equations

Physically Implement: AND-OR and NOR-NOR PLAs

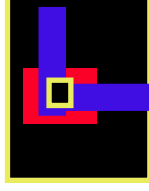
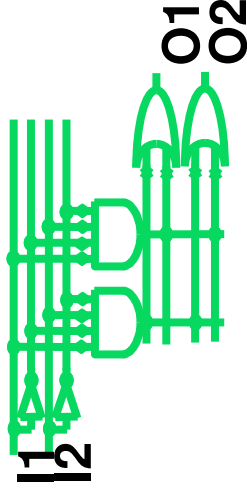
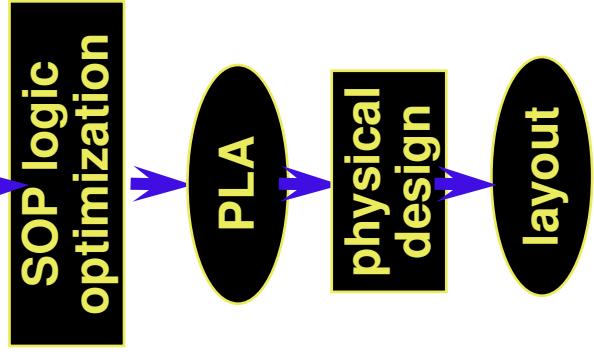


Logic increases with the number of product terms

Early “Synthesis” Flow



$$F1 = B + D + A C + A C$$



Key Technology: SOP Logic Minimization

Can realize an arbitrary logic function in sum-of-products or two-level form

$$F1 = \bar{A} \bar{B} + \bar{A} B D + \bar{A} B \bar{C} \bar{D} \\ + A B C \bar{D} + A \bar{B} + A B D$$

$$F1 = \bar{B} + D + \bar{A} \bar{C} + A C$$

Of great interest to find a minimum sum-of-products representation

Definitions - 1

Basic definitions:

Let $B = \{0, 1\}$ and $Y = \{0, 1, 2\}$

→ don't care – aka “X”

Input variables: $X_1, X_2 \dots X_n$

Output variables: $Y_1, Y_2 \dots Y_m$

A logic function **ff** (or Boolean function, switching function) in **n** inputs and **m** outputs is the map

ff: $B^n \longrightarrow Y^m$

Definitions - 2

If $b \in B^n$ is mapped to a 2 then function is incompletely specified, else completely specified

For each output we define:

ON-SET_i $\subseteq B^n$, the set of all input values for which $ff_i(x) = 1$

OFF-SET_i $\subseteq B^n$, the set of all input values for which $ff_i(x) = 0$

DC-SET_i $\subseteq B^n$, the set of all input values for which $ff_i(x) = 2$

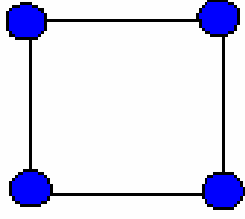
The Boolean n-Cube, B^n



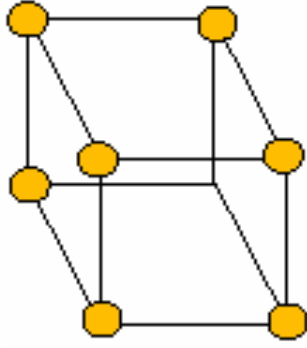
B^0



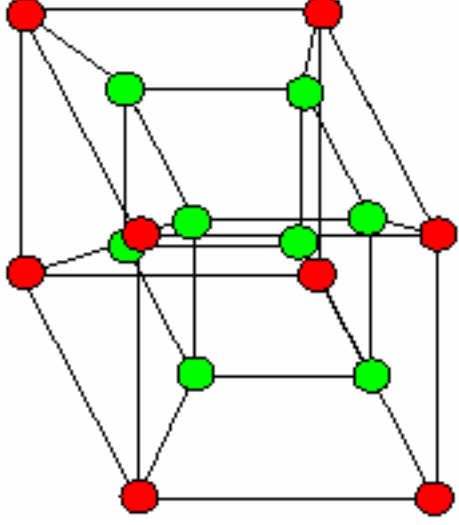
B^1



B^2



B^3



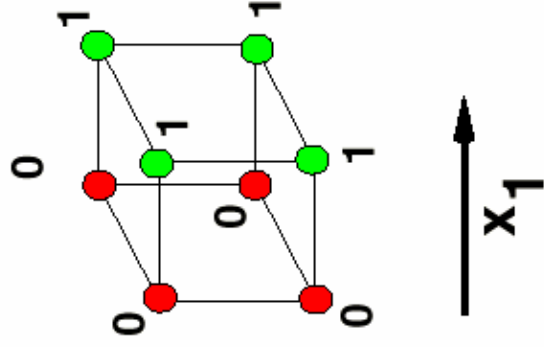
B^4

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

Literals

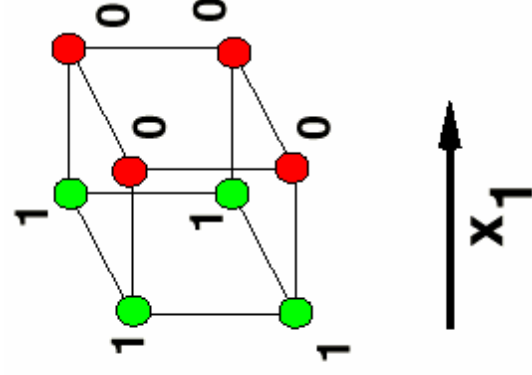
A literal is a variable or its negation y, \bar{y}

It represents a **logic function**



$$f = x_1$$

Green – ON-set
Red – OFF-set



$$f = \bar{x}_1$$

Boolean Formulas -- Syntax

Boolean functions can be represented by formulas defined as catenations of

- parentheses - (,)
- literals - $x, y, z, \bar{x}, \bar{y}, \bar{z}$
- Boolean operators - + (OR), \times (AND)
- complementation - e.g. $\overline{x + y}$

$$\begin{aligned} \text{Examples: } f &= x_1 \times \bar{x}_2 + \bar{x}_1 \times x_2 \\ &= (x_1 + x_2) \times (\bar{x}_1 + \bar{x}_2) \\ h &= a + b \times c \\ &= \overline{\bar{a} \times (\bar{b} + \bar{c})} \end{aligned}$$

We will usually replace \times by catenation, e.g. $a \times b \rightarrow ab$.

“Semantic” Description of Boolean Function

EXAMPLE: Truth table form of an incompletely specified function

ff: $B^3 \xrightarrow{\text{red arrow}} Y^2$

X_1	X_2	X_3	Y_1	Y_2
0	0	0	1	1
0	0	1	1	0
0	1	0	0	1
0	1	1	0	1
1	0	0	1	0
1	0	1	1	2
1	1	0	1	1
1	1	1	2	1

Y_1 : ON-SET₁ = {000, 001, 100, 101, 110}
OFF-SET₁ = {010, 011}
DC-SET₁ = {111}

Cube Representation

$$F1 = \bar{A}\bar{B} + \bar{A}BD + \bar{A}\bar{B}\bar{C}\bar{D} \\ + ABC\bar{D} + AB + ABD$$

Inputs	Outputs
00--	1
01-1	1
0100	1
1110	1
10--	1
11-1	1

$$F1 = \bar{B} + D + \bar{A}\bar{C} + AC$$



minimum representation

-0--	1
---1	1
0-0-	1
1-1-	1

Operations on Logic Functions

- (1) Complement: $f \longrightarrow \bar{f}$
interchange ON and OFF-SETS
- (2) Product (or intersection or logical AND)
 $h = f \bullet g$ or $h = f \cap g$
- (3) Sum (or union or logical OR):
 $h = f + g$ or $h = f \cup g$
- (4) Difference $h = f - g = f \cap \bar{g}$

Prime Implicants

A cube p is an implicant of f if it does not intersect the OFF-SET of f

$$p \subseteq f_{\text{ON}} \cup f_{\text{DC}} \text{ (or } p \cap f_{\text{OFF}} = 0\text{)}$$

A prime implicant of f is an implicant p such that

- (1) No other implicant q is such that $q \supset p$ in the sense that q covers all vertices of p
- (2) $f_{\text{DC}} \not\supset p$

A minterm is a fully specified implicant
e.g., **011**, **111** (not **01-**)

Examples of Implicants/Primes

X_1	X_2	X_3	Y_1
0	0	0	1
0	0	1	1
0	1	0	0
0	1	1	0
1	0	0	1
1	0	1	1
1	1	0	1
1	1	1	2

000, 00- are implicants, but not primes (-0-)

1-1

0-0

Prime and Irredundant Covers

A cover is a set of cubes C such that
 $C \supseteq f_{ON}$ and
 $C \subseteq f_{ON} \cup f_{DC}$

All of the ON-set is covered by C

C is contained in the ON-set and Don't Care Set

A prime cover is a cover whose cubes are all prime implicants

An irredundant cover is a cover C such that removing any cube from C results in a set of cubes that no longer covers the function

Minimum covers

A minimum cover is a cover of minimum cardinality

Theorem: A minimum cover can always be found by restricting the search to prime and irredundant covers.

Given any cover **C**

- (a) if redundant, not minimum
- (b) if any cube **q** is not prime, replace **q** with prime **p** \supset **q** and continue until all cubes prime; it is a minimum prime cover

Example Covers

X_1	X_2	X_3	Y_1
0	0	0	1
0	0	1	1
0	1	0	0
0	1	1	0
1	0	0	1
1	0	1	1
1	1	0	1
1	1	1	2

00-

10- is a cover. Is it prime?

11- Is it irredundant?

What is a minimum prime and irredundant cover for the function?

Example Covers

X_1	X_2	X_3	Y_1
0	0	0	1
0	0	1	1
0	1	0	0
0	1	1	0
1	0	0	1
1	0	1	1
1	1	0	1
1	1	1	2

00-

is a cover. Is it prime?
Is it irredundant?

10-

11-

-0-

is a cover. Is it prime?
Is it irredundant?
Is it minimum?

11-

What is a minimum prime and irredundant cover for the function?

The Quine - McCluskey Method

- Step 1:** List all minterms in ON-SET and DC-SET
- Step 2:** Use a prescribed sequence of steps to find all the prime implicants of the function
- Step 3:** Construct the prime implicant table
- Step 4:** Find a minimum set of prime implicants that cover all the minterms

Example

0	0000	0,8	-000	Ⓔ	8,9,10,11	10--	Ⓑ
5	0101	5,7	01-1	Ⓓ	10,11,14,15	1-1-	Ⓐ
7	0111	7,15	-111	Ⓒ			
8	1000	8,9	100-				
9	1001	8,10	10-0				
10	1010	9,11	10-1				
11	1011	10,11	101-				
14	1110	10,14	1-10				
15	1111	11,15	1-11				
		14,15	111-				

Ⓐ Ⓑ Ⓒ Ⓓ Ⓔ are prime implicants

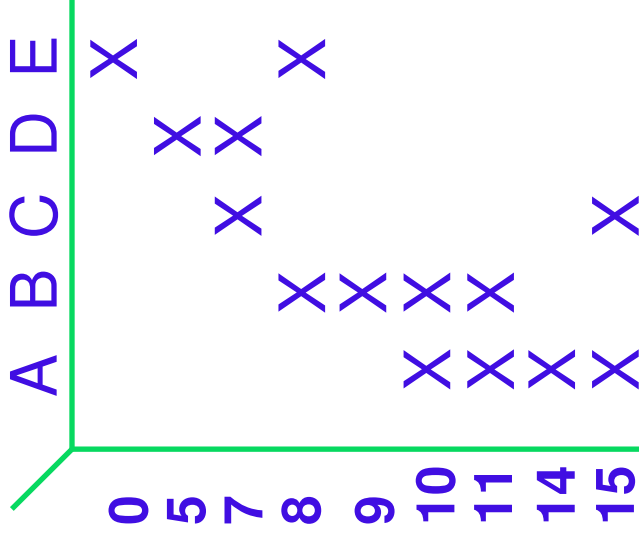
Prime Implicant Table

Minterms
(ON-SET only)

	A	B	C	D	E
0					X
5		X			
7		X	X		
8			X		X
9		X			
10	X	X			
11	X	X			
14	X				
15	X				X

X's indicate minterms covered by PIs

Essential Prime Implicants



Row with a single X identifies an essential prime implicant (EPI)

Essential PI's E, D, B, A \Rightarrow Form minimum cover

Dominating Rows

In general EPIs do not form a cover

At Step 4, we need to select PIs to add to the EPIs so as to form a minimum cover

	A	B	C	D	F	G
1	X	X				
8	X		X			
9	X	X	X			
24	X					X
25	X		X		X	X
27					X	X

Row 9 dominates 8

Row 25 dominates 24

Can remove 8 since covering 9 implies covering of 8

Dominating Columns

	A	B	C	D	F	G
1	X	X				
8	X		X			
9	X	X	X			
24	X					X
25	X		X		X	X
27					X	X

F dominates **D**

Can remove **D** since **F** covers all minterms **D** covers

Can this happen in the original table?

May happen after removal of PIs

Step 4 Issues

Removal of dominating columns or dominated rows may introduce columns with single X 's.

- **Need to iterate**

A cover may still not be formed after all essential elements and dominance relations have been removed

- **Need to branch over possible solutions**

Recursive Branching (Step 4)

- (a) Select EPIs, remove dominated columns and dominating rows iteratively till table does not change**
- (b) If the size of the selected set (+ lower bound) exceeds or equals best solution so far, return from this level of recursion. If no elements left to be covered, declare selected set as the best solution recorded.**
- (c) Select (heuristically) a branching column.**

Recursive Branching (Step 4) - 2

- (d) Given the selected column, recur**
 - On the sub-table resulting from deleting the column and all rows covered by this column. Add this column to the selected set.**
 - On the sub-table resulting from deleting the column without adding it to the selected set.**

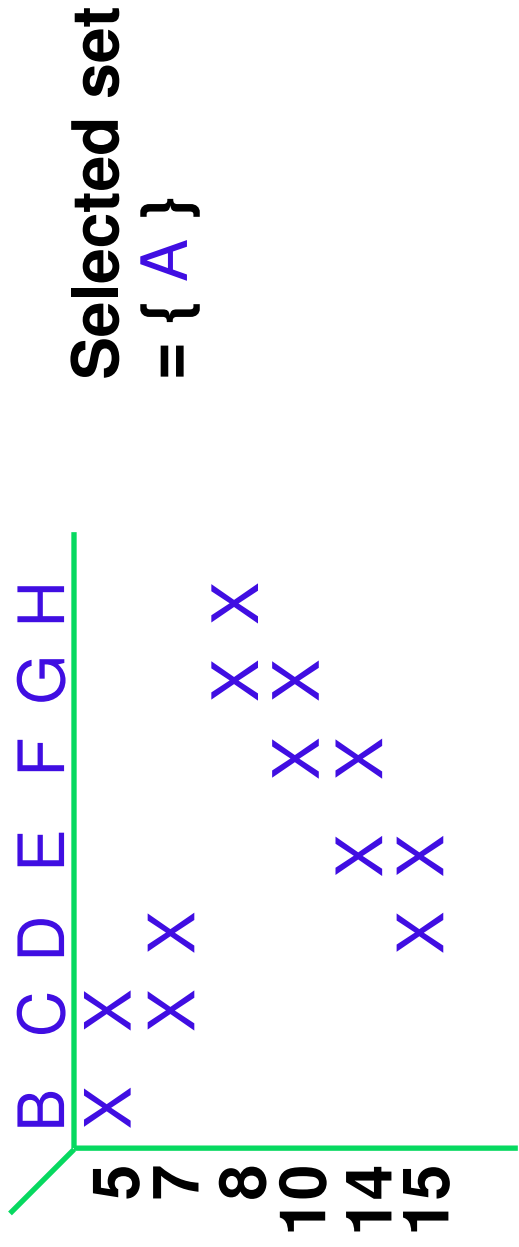
Example - a1

	A	B	C	D	E	F	G	H
0	X							X
1	X	X						
5		X	X					
7		X	X					
8							X	X
10						X	X	
14					X	X		
15				X	X			

No essential primes, dominated rows or columns.

Select prime A

Example - a2



B is dominated by C

H is dominated by G

Remove B, H

Example - a3

	C	D	E	F	G
5	X				
7	X	X			
8					X
10				X	X
14				X	X
15				X	X

C, G essential to
this table

Selected set
= {A, C, G}

	D	E	F
14	X	X	X
15	X	X	

Selected set
= {A, C, G, E}

Example - b1

	B	C	D	E	F	G	H
0							
1	X						X
5	X	X					
7	X	X					
8				X	X		
10				X	X		
14			X	X			
15			X	X			

Selected set = { }

Essential primes
in this table are B, H

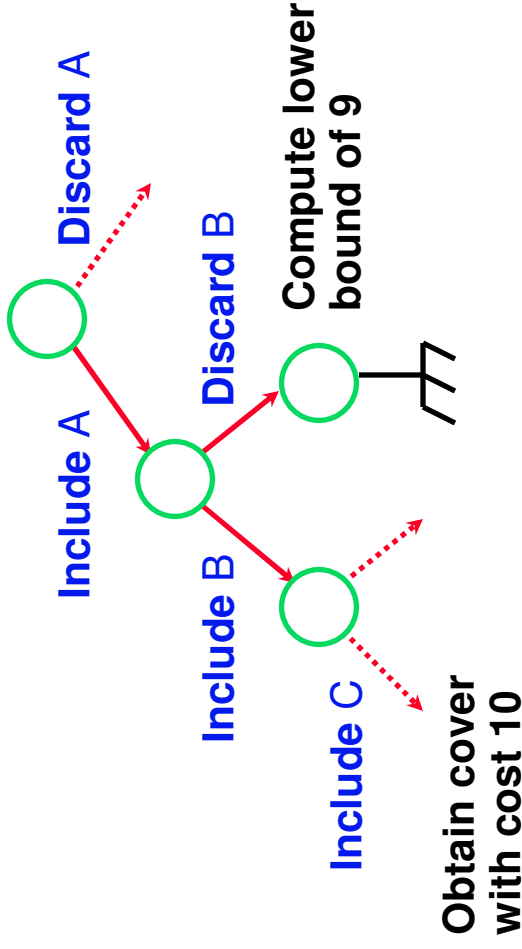
Selected set = {B, H}

	C	D	E	F	G
7	X	X			
10				X	X
14				X	X
15			X	X	

Selected set
= {B, H, D, F}

Espresso-Exact (1987)

Efficient lower bounding at Step 4(b) to terminate unprofitable searches high in the recursion



Size of selected set + Lower bound equals or exceeds best solution already known, quit level of recursion

Lower Bounding

	A	B	C	D	E	F
0	X				X	
1	X		X			X
4					X	
6		X	X			
8		X				
10				X	X	
12				X		X

Lower bound: Maximal independent set of rows all of which are pairwise disjoint

Maximal independent set = {1, 4, 8} or {0, 6, 10}

Need to select at least one PI/column to cover each row.

NOTE: Finding maximum independent set is itself NP-hard

Complexity of Q-M based Methods

- (1) There exist functions for which the number of prime implicants is $O(3^n)$ (n is number of inputs)
- (2) Given a PI table, recursive branching could require $O(2^m)$ time (m is the number of PIs)

Current logic minimizers able to find exact solutions for functions with 20-25 input variables

⇒ Need heuristic methods for larger functions

Heuristic Logic Minimization

Presently, there appears to be a limit of ~20-25 input variables in problems that can be handled by exact minimizers

Easy for complex control logic to exceed 20- 25 input variables

HISTORY

50's	Karnaugh Map	≤ 5 variables
60's	Q-M method	< 10 variables
70's	Starnner, Dietmeyer	< 15 variables
1974	MINI	heuristic
1980-84	ESPRESSO	approaches
1986	McBoole	< 25 variables
1987	ESPRESSO-EXACT	< 25 variables

Also, Multiple Output Functions

Truth table is AND-OR representation

AND			OR	
a	b	c	f	g
0	1	–	1	0
0	1	1	1	1
1	0	1	0	1

What does vector **0 1 1** produce?

ON-SET of **f** = {**0 1 –**, **0 1 1**} = {**0 1 –**}

ON-SET of **g** = {**0 1 1**, **1 0 1**}

Multiple-Output Function Primes

Same definition as in single-output case

- Cube with most minterms that will intersect OFF-SET if you add any more minterms to them

<u>f g</u>	<u>CUBE</u>	<u>TYPE</u>
0000	0000	10
0001	000–	10
1001	1001	10
0000	1001	11
0010	000–	11
1001	000–	11

MINI

S.J. Hong, R.G. Cain, D.L. Ostapko - 1974

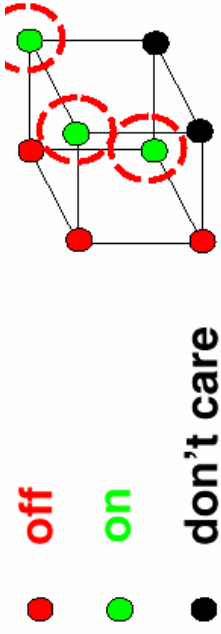
Final solution is obtained from initial solution by iterative improvement rather than by generating and covering prime implicants

Three basic modifications are performed

- Reduction of implicants while maintaining coverage**
- Reshaping implicants in pairs**
- Expansion of implicants (and removal of covered implicants)**

Example: Expansion

Consider $\mathcal{F}(a, b, c) = (f, d, r)$, where $f = \{\bar{a}b\bar{c}, a\bar{b}c, abc\}$ and $d = \{a\bar{b}\bar{c}, abc\}$, and the sequence of covers illustrated below:



$$F^1 = abc + \bar{a}bc + abc\bar{b}$$

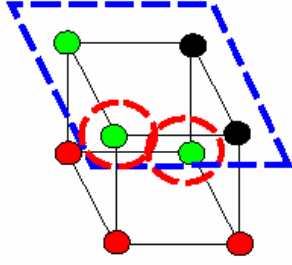
EXPAND $abc \rightarrow a$



$$F^2 = a + \bar{a}bc + \bar{a}bc\bar{b}$$

$\bar{a}bc$ is redundant
 a is prime

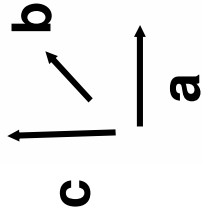
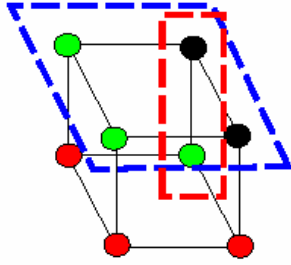
$$F^3 = a + \bar{a}bc$$



EXPAND $\bar{a}bc \rightarrow \bar{b}c$



$$F^4 = a + \bar{b}c$$



Reduction

Reduce the size (in the sense of the number of minterms/vertices that it covers) of cubes in f without affecting coverage

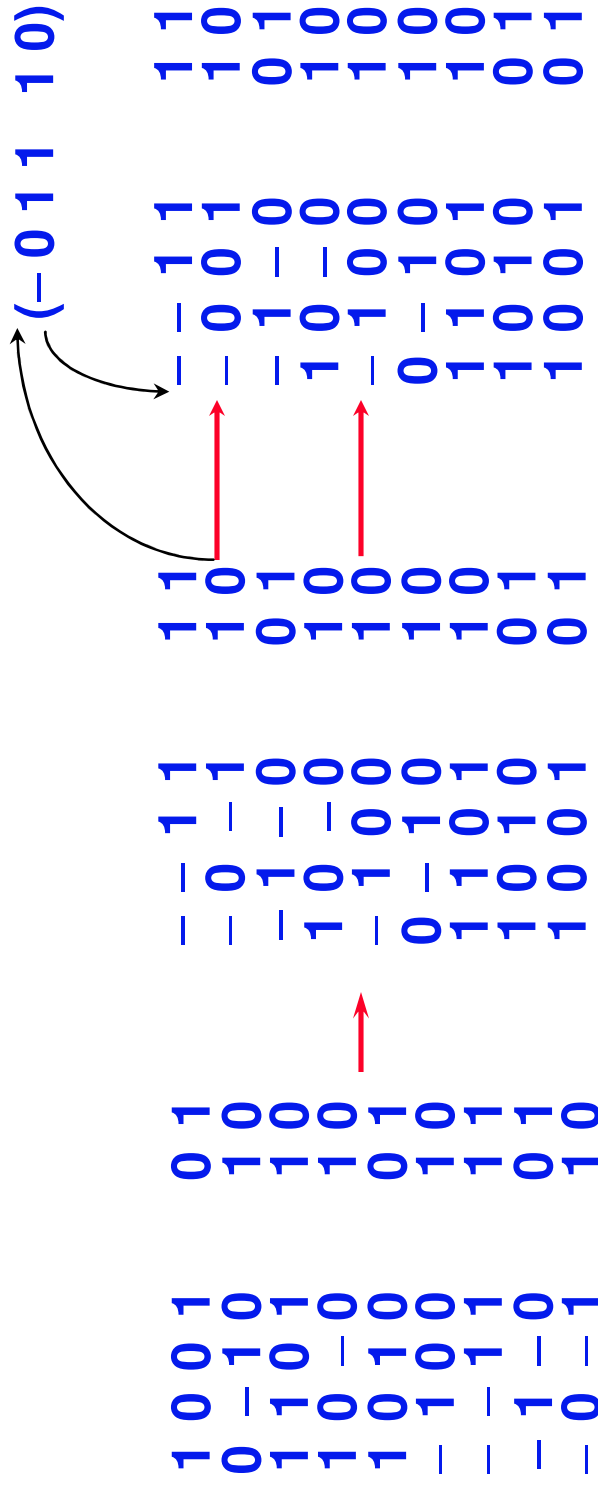
The smaller the size of the cube, the more likely it will be covered by an expanded cube

Reduction Examples

Reducing covers:

$$\begin{array}{cccc}
 \mathbf{f} & 1 & - & 1 \\
 & - & 1 & - \\
 & - & - & 1 \\
 & & & 1
 \end{array}
 \qquad
 \begin{array}{cccc}
 & & & 1 \\
 & & & - \\
 & & & - \\
 & & & 1
 \end{array}
 \qquad
 \begin{array}{cccc}
 & & & 1 \\
 & & & - \\
 & & & - \\
 & & & 1
 \end{array}
 \qquad
 \begin{array}{cccc}
 & & & 1 \\
 & & & - \\
 & & & - \\
 & & & 1
 \end{array}$$

$\mathbf{f}_{\text{reduced}}$



Reorder,
put larger cubes first

Reshaping

Attempt to change the shape of the cubes without changing coverage or number

Reshaping transforms a pair of cubes into another pair such that coverage is unaffected (perturbs solution so next expand does things differently)

Reshaping Example

	f	f_{ordered}	
	<pre> 1 1 - 0 1 - 1 - 1 0 - - 1 0 0 0 1 0 1 1 0 1 0 1 1 0 0 1</pre>	<pre> 1 1 1 0 0 1 1 0 1 0 1 0 0 1 0 1</pre>	<pre> 1 1 1 0 1 0 1 0 1 0 1 0 0 1 0 1</pre>

<pre> 1 1 - 1 0 - 1 0 0 1 - - 1 0 0 0 1 0 1 1 0 1 0 0 1</pre>	<pre> 1 1 - 1 0 - 1 0 0 1 0 0 1 0 0 0 0 0 1 0 0 0 1 0 0 - 1 1</pre>	<pre> 1 1 0 1 1 1 1 0 1 1 1 0 1 1 1 0 1 0</pre>	<pre> 1 1 0 1 1 1 1 0 1 1 1 0 1 1 1 0 1 0</pre>
<pre> 1 (2,5) (3,8) (4,9) 6 7</pre>		f_{reshaped}	

A Complete Example

		f				g			
		00	01	11	10	00	01	11	10
ab	00	9		10		5	5		00
	01	4		4				2	01
cd	11	1	1	1	1	1	1	1	11
	10	8	7	8		5	5	6	10

initial f

1	2	3	4	5	6	7	8	9	10
a	b	c	d	f	g				
-	1	0	1	1	1	-	1	1	1
1	0	1	1	0	1	1	0	1	0
-	1	0	1	1	0	1	1	0	1
1	0	1	1	0	1	1	0	1	0
0	1	1	0	1	0	1	0	1	0
-	1	0	0	1	1	0	1	1	0
1	0	0	0	1	1	0	1	1	0

↑ expand ↑ ↑ ↑

↓ ↓

Example - 2

	ab		00	01	11	10	00	01	11	10	
cd	00	01	10	11	10	00	01	11	10	00	
	00	9	10	9	10		5	5		01	
	01	4	4	3	4				2	11	
	11	1,4	1,4	1	1,4	1	1	1	1	10	
	10	7	10	7	10		5	5	6		

expanded f

	a	b	c	d	f	g
1	-	1	0	1	1	1
2	1	0	1	1	0	1
3	1	1	0	1	1	0
4	-	1	0	1	1	0
5	-	1	0	1	0	1
6	1	0	1	0	0	1
7	0	-	1	0	1	0
9	-	1	0	1	1	0
10	1	0	-	1	1	0

reduce
→

	-	1	0	1	1	1
	1	0	1	1	0	1
	1	1	0	1	1	0
	-	1	0	1	1	0
	-	1	0	1	0	1
	1	0	1	0	0	1
	0	-	1	0	1	0
	-	1	0	1	1	0

Example - 3

ab		cd				reduced f			
		00	01	11	10	00	01	11	10
00		9		10		5		5	
01	4		3	4				2	
11	1	1	1	1	1	1	1	1	1
10	7	7		10		5	5	6	6

	a	b	c	d	f	g	
1	-	1	1	1	1	1	2,4
2	1	0	0	1	0	1	2,4
3	1	1	0	1	1	0	5,9
4	-	1	0	0	0	1	6,10
5	-	1	0	1	0	1	5,9
6	1	0	1	0	1	1	6,10
7	0	-	1	0	0	1	
9	-	1	0	0	-	1	
10	1	0	-	1	1	0	

reshape

Example - 5

ab		cd				fg			
		00	01	11	10	00	01	11	10
00		9	3,9	3			5,9	5,9	
01	4		3	2,3,4					2
11	1,4,7	1,7	1	1,2,4,6	1	1	1	1	1,2,6
10	7	7		6			5	5	6

final
expanded f

		abcd				fg			
1		-	1	1		1	1	1	
2		1	0	-	1	1	0	1	0
3		1	0	-	1	0	1	0	1
4		-	1	0	-	0	1	1	0
5		-	1	0	-	0	1	1	0
6		1	0	1	0	1	1	0	1
7		0	-	1	0	1	1	0	1
9		-	1	0	0	1	1	0	1

final F

Summary of 2-level

2-level optimization is very effective and mature.

Espresso (developed at Berkeley) is the “last word” on the subject

2-level optimization is directly useful for

**PLA’s/PLD’s – these were widely used to implement complex control logic in the early 80’s
– they are rarely used these days**

2-level optimization forms the theoretical foundation for multilevel logic optimization

2-level optimization is useful as a subroutine in multilevel optimization