

Two-level Logic Synthesis and Optimization

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Module 1

□ Objectives

- Fundamentals of logic synthesis
- Mathematical formulation
- Definition of the problems

Combinational logic design

Background

□ Boolean Algebra

- Quintuple $(B, +, \cdot, 0, 1)$
- Binary Boolean algebra $B = \{0, 1\}$

□ Boolean function

- Single output $f : B^n \rightarrow B$
- Multiple output $f : B^n \rightarrow B^m$
- Incompletely-specified:
 - *Don't care* symbol: $*$
 - $f : B^n \rightarrow \{0, 1, *\}^m$

The *don't care* conditions

- We do not care about the value of a function
- Related to the environment
 - Input patterns that never occur
 - Input patterns such that some output is never observed
- Very important for synthesis and optimization

Definitions

□ Scalar function:

□ ON-set

- Subset of the domain such that **f** is true

□ OFF-set

- Subset of the domain such that **f** is false

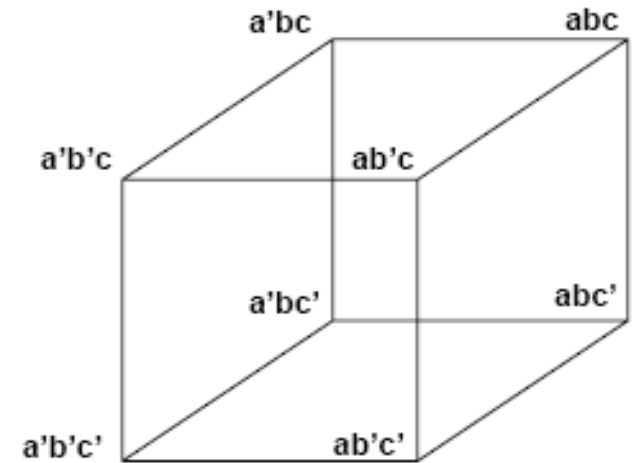
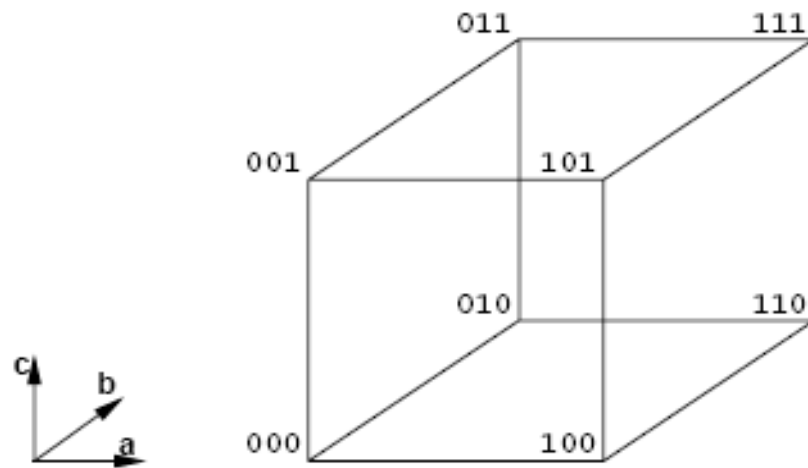
□ DC-set

- Subset of the domain such that **f** is a *don't care*

□ Multiple-output function:

- ON, OFF, DC-sets defined for each component

Cubical representation



Definitions

- **Boolean variables**
- **Boolean literals:**
 - Variables and their complement
- **Product or cube:**
 - Product of literals
- **Implicant:**
 - Product implying a value of the function (usually 1)
 - Hypercube in the Boolean space
- **Minterm:**
 - Product of all input variables implying a value of the function (usually 1)
 - Vertex in the Boolean space

Tabular representations

□ Truth table

- List of all minterms of a function

□ Implicant table or cover

- List of implicants sufficient to define a function

□ Note

- Implicant tables are smaller in size as compared to truth tables

Example of truth table

$$\square x = ab + a'c; \quad y = ab + bc + ac$$

abc	xy
000	00
001	10
010	00
011	11
100	00
101	01
110	11
111	11

Example of implicant table

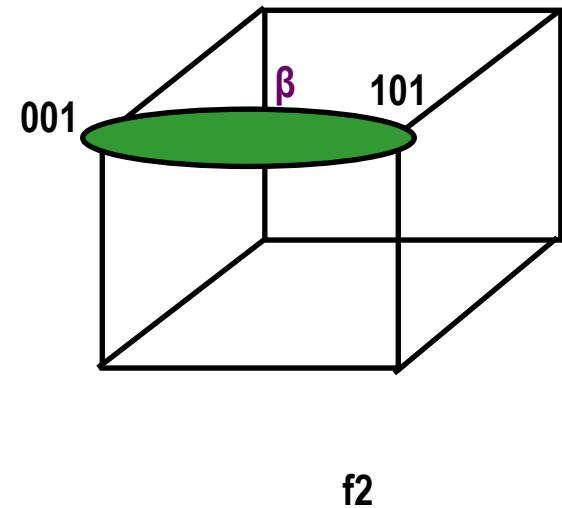
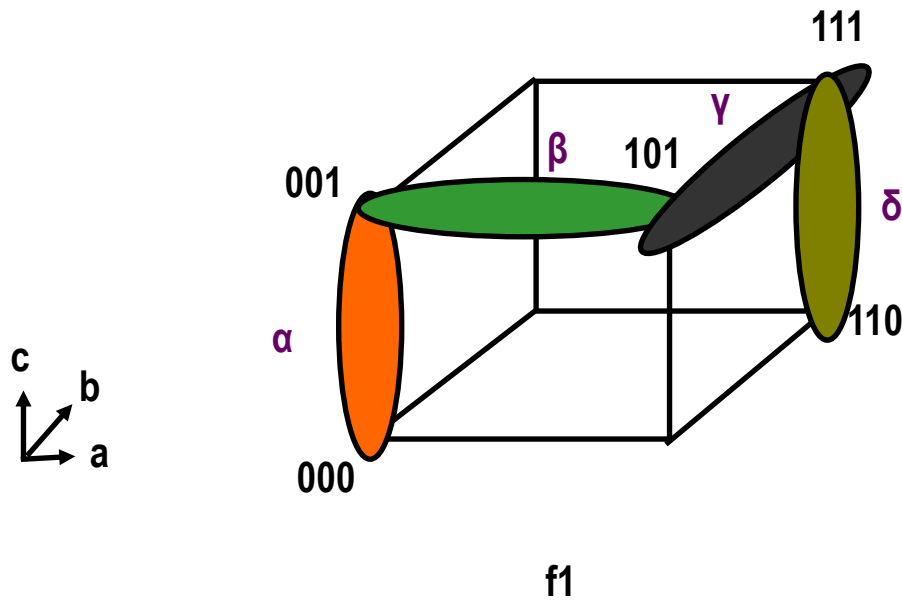
$$\square x = ab + a'c; \quad y = ab + bc + ac$$

abc	xy
001	10
*11	11
101	01
11*	11

Cubical representation of minterms and implicants

$$\square f_1 = a'b'c' + a'b'c + ab'c + abc + abc'$$

$$\square f_2 = a'b'c + ab'c$$



Representations

□ Visual representations

- Cubical notation
- Karnaugh maps

□ Computer-oriented representations

□ Matrices

- Sparse
- Various encoding

□ Binary-decision diagrams

- Address sparsity and efficiency

Module 2

□ Objectives

- Two-level logic optimization
- Motivation
- Models
- Exact algorithms for logic optimization

Two-level logic optimization motivation

- Reduce size of the representation
- Direct implementation
 - PLAs reduce size and delay
- Other implementation styles
 - Reduce amount of information
 - Simplify local functions and connections

Programmable logic arrays

□ Macro-cells with rectangular structure

- Implement any multi-output function
- Layout generated by module generators
- Fairly popular in the seventies/eighties

□ Advantages

- Simple, predictable timing

□ Disadvantages

- Less flexible than cell-based realization
- Dynamic operation

□ Open issue

- Will PLA structures be useful with new nanotechnologies? (e.g., nanowires)

Programmable logic array

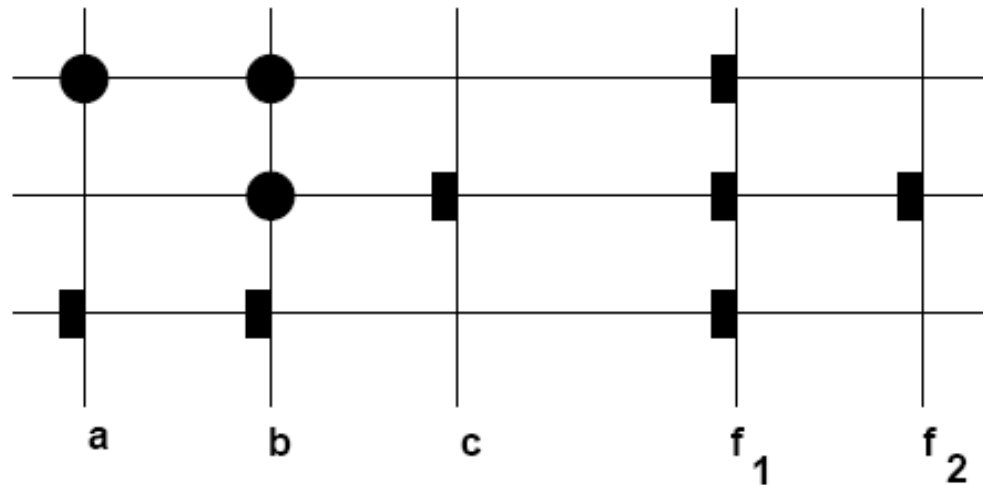
$$\square f_1 = a'b' + b'c + ab; \quad f_2 = b'c$$

00~~X~~ 10

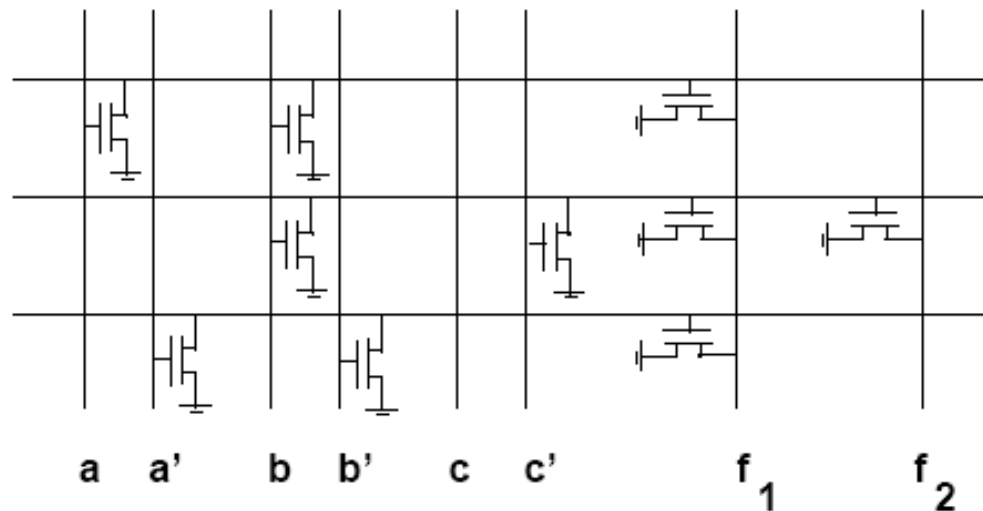
~~X~~01 11

11~~X~~ 10

(a)



(b)



(c)

(c)

Two-level minimization

□ Assumptions

- Primary goal is to reduce the number of implicants
- All implicants have the same cost
- Secondary goal is to reduce the number of literals

□ Rationale

- Implicants correspond to PLA rows
- Literals correspond to transistors

Definitions

□ **Minimum cover**

- Cover of a function with minimum number of implicants
- Global optimum

□ **Minimal cover or irredundant cover**

- Cover of the function that is not a proper superset of another cover
- No implicant can be dropped
- Local optimum

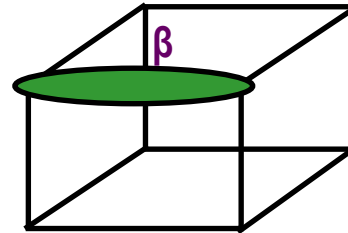
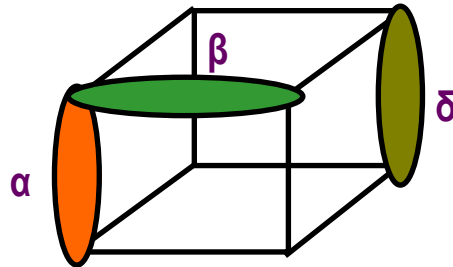
□ **Minimal w.r.to 1-implicant containment**

- No implicant contained by another one
- Weak local optimum

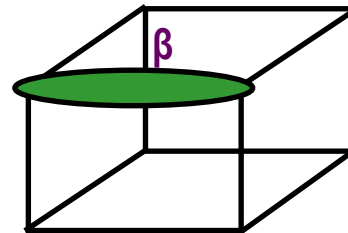
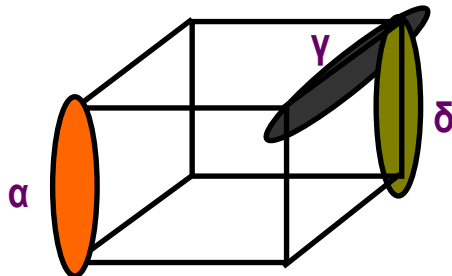
Example

□ $f_1 = a'b'c' + a'b'c + ab'c + abc + abc'$; $f_2 = a'b'c + ab'c$

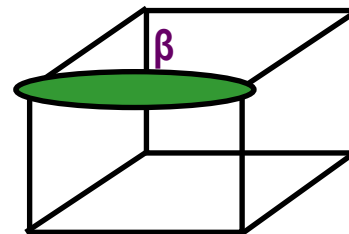
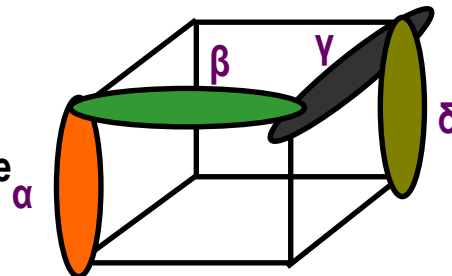
Minimum cover



Irredundant cover



Minimal cover w.r. to single implicant containment



f1

f2

Definitions

□ Prime implicant

- Implicant not contained by any other implicant

□ Prime cover

- Cover of prime implicants

□ Essential prime implicant

- There exist some minterm covered only by that prime implicant
- Needs to be included in the cover

Two-level logic minimization

□ Exact methods

- Compute minimum cover
- Often difficult/impossible for large functions
- Based on Quine-McCluskey method

□ Heuristic methods

- Compute minimal covers (possibly minimum)
- Large variety of methods and programs
 - MINI, PRESTO, ESPRESSO

Exact logic minimization

- **Quine's theorem:**

- There is a minimum cover that is prime

- **Consequence**

- Search for minimum cover can be restricted to prime implicants

- **Quine-McCluskey method**

- Compute prime implicants
 - Determine minimum cover

Prime implicant table

- Rows: minterms
- Columns: prime implicants
- Exponential size
 - 2^n minterms
 - Up to $3^n / n$ prime implicants
- Remarks
 - Some functions have much fewer primes
 - Minterms can be grouped together
 - Implicit methods for implicant enumeration

Example

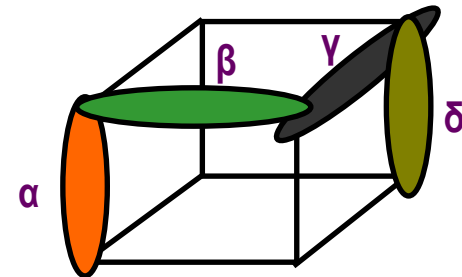
□ $f = a'b'c' + a'b'c + ab'c + abc + abc'$

□ Primes:

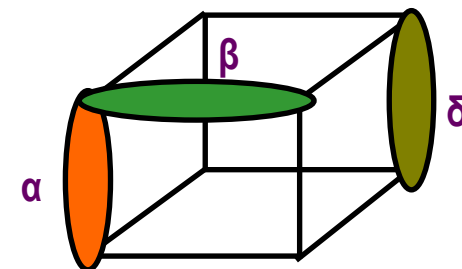
α	00*	1
β	*01	1
γ	1*1	1
δ	11*	1

□ Table:

	α	β	γ	δ
000	1	0	0	0
001	1	1	0	0
101	0	1	1	0
111	0	0	1	1
110	0	0	0	1



Prime implicants of f



Minimum cover of f

Minimum cover early methods

□ Reduce table

- Iteratively identify essentials,
save them in the cover.
Remove covered minterms

□ Petrick's method

- Write covering clauses in *pos* form
- Multiply out pos form into *sop* form
- Select cube of minimum size

□ Remark

- Multiplying out clauses has exponential cost

Example

- **pos clauses**

- $(\alpha) (\alpha + \beta) (\beta + \gamma) (\gamma + \delta) (\delta) = 1$

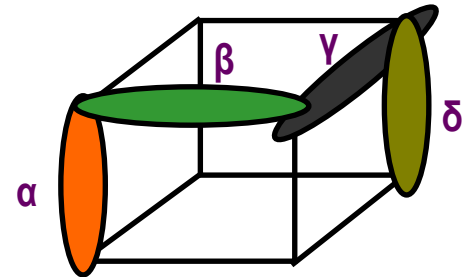
- **sop form:**

- $\alpha\beta\delta + \alpha\gamma\delta = 1$

- **Solutions:**

- $\{ \alpha \beta \delta \}$

- $\{ \alpha \gamma \delta \}$



Matrix representation

- View table as Boolean matrix: A
- Selection Boolean vector for primes: x
- Determine X such that
 - $Ax \geq 1$
 - Select enough columns to cover all rows
- Minimize cardinality of x

Example

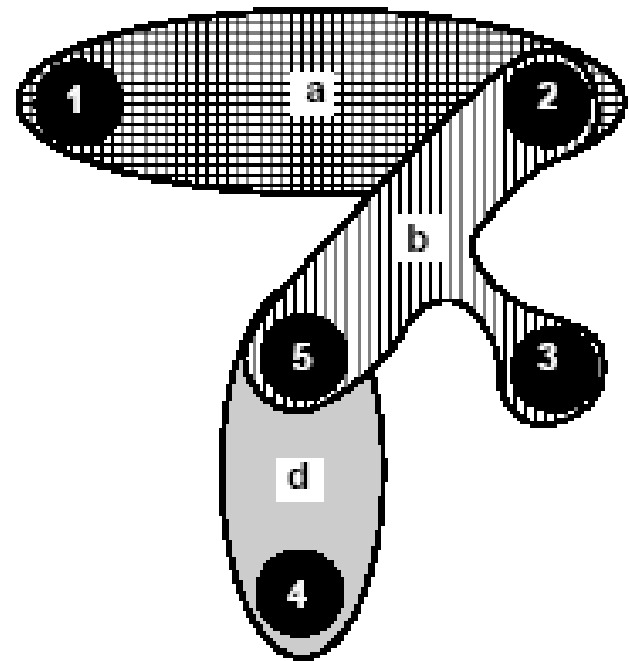
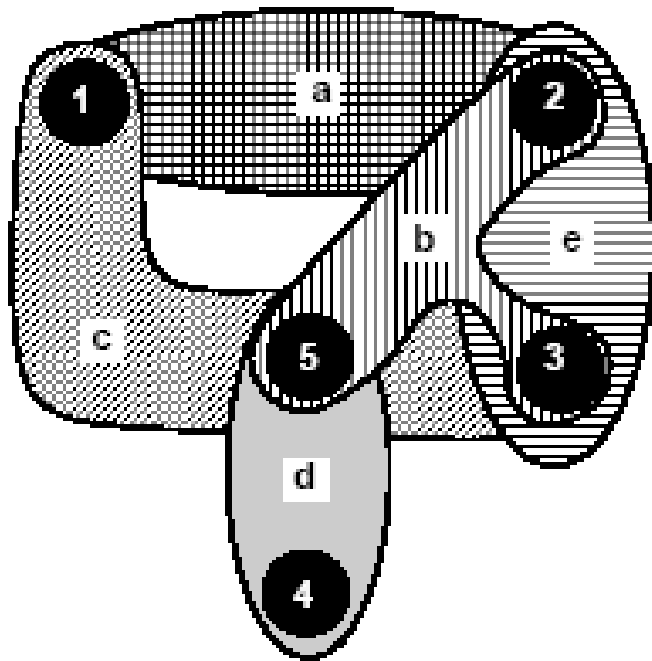
$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

Covering problem

- **Set covering problem:**
 - A set **S** -- minterm set
 - A collection **C** of subsets (implicant set)
 - Select fewest elements of **C** to cover **S**
- **Computationally intractable problem**
- **Exact solution method**
 - Branch and bound algorithm
- **Several heuristic approximation methods**

Example

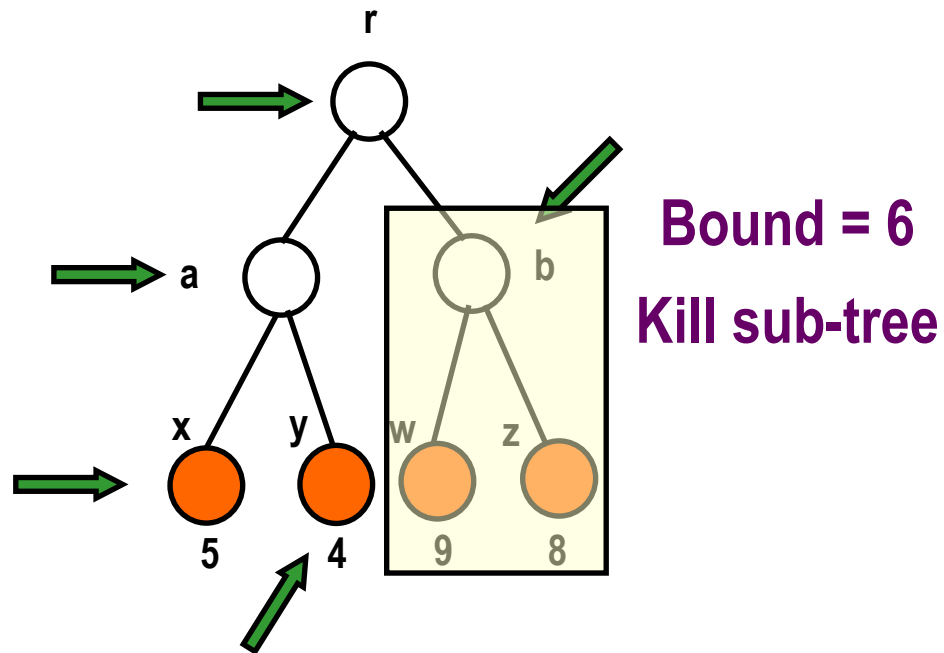
Edge-cover of a hypergraph



Branch and bound algorithm

- **Tree search in the solution space**
 - **Potentially exponential**
- **Use bounding function:**
 - **If the lower bound on the solution cost that can be derived from a set of future choices exceeds the cost of the best solution seen so far, then kill the search**
 - **Bounding function should be fast to evaluate and accurate**
- **Good pruning may expedite the search**

Example



Branch and bound for logic minimization

Reduction strategies

- Use matrix formulation of the problem
- Partitioning:
 - If **A** is block diagonal:
 - Solve covering problems for the corresponding blocks
- Essentials
 - Column incident to one (or more) rows with single 1
 - Select column
 - Remove covered row(s) from table

Branch and bound for logic minimization

Reduction strategies

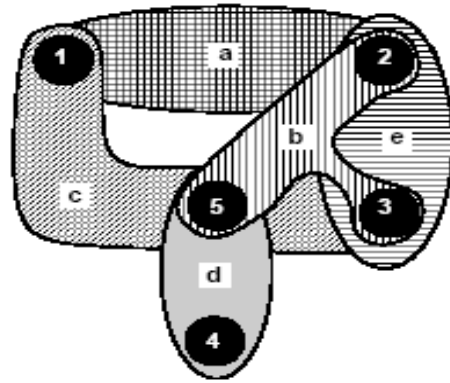
□ Column (implicant) dominance:

- If $a_{ki} \geq a_{kj}$ for all k
 - Remove column j (dominated)
- Dominated implicant (j) has its minterms already covered by dominant implicant (i)

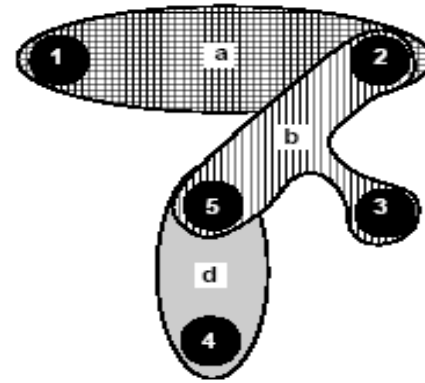
□ Row (minterm) dominance:

- If $a_{ik} \geq a_{jk}$ for all k
 - Remove row i (dominant)
- When an implicant covers the dominated minterm, it also covers the dominant one

Example



(a)



(b)

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 \end{bmatrix}$$

Example

- Fourth column is essential
- Fifth column is dominated
- Fifth row is dominant
- Matrix after reductions:

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

Branch and bound covering algorithm

```
EXACT_COVER( $A, x, b$ ) {  
    Reduce matrix  $A$  and update corresponding  $x$ ;  
    if (current_estimate  $\geq |b|$ ) return ( $b$ );  
    if ( $A$  has no rows) return( $x$ );  
    select a branching column  $c$ ;  
     $x_c = 1$ ;  
     $\tilde{A} = A$  after deleting  $c$  and rows incident to it;  
     $\tilde{x} = EXACT\_COVER(\tilde{A}, x, b)$ ;  
    if (  $|\tilde{x}| < |b|$  )  
         $b = \tilde{x}$ ;  
     $x_c = 0$ ;  
     $\tilde{A} = A$  after deleting  $c$ ;  
     $\tilde{x} = EXACT\_COVER(\tilde{A}, x, b)$ ;  
    if (  $|\tilde{x}| < |b|$  )  
         $b = \tilde{x}$ ;  
    return( $b$ );  
}
```

Bounding function

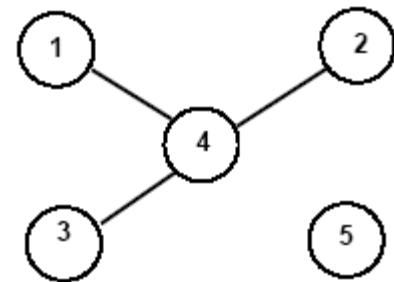
- Estimate lower bound on covers that can be derived from current solution vector \mathbf{x}
- The sum of the 1s in \mathbf{x} , plus bound of cover for local \mathbf{A}
 - Independent set of rows
 - No 1 in the same column
 - Require independent implicants to cover
 - Construct graph to show pairwise independence
 - Find clique number
 - Size of the largest clique
 - Approximation (lower) is acceptable

Example

□ Row 4 independent from 1,2,3

□ Clique number and bound is 2

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 \end{bmatrix}$$



Example

□ There are no independent rows

□ Clique number is 1 (one vertex)

□ Bound is $1+1=2$

□ Because of the essential already selected

$$A = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

Example

Branching on the cyclic core

□ Select first column

- Recur with $\tilde{A} = [11]$
 - Delete one dominated column
 - Take other column (essential)
- New cost is 3

$$A = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

□ Exclude first column

- Find another solution with cost equal to 3.
- Discard

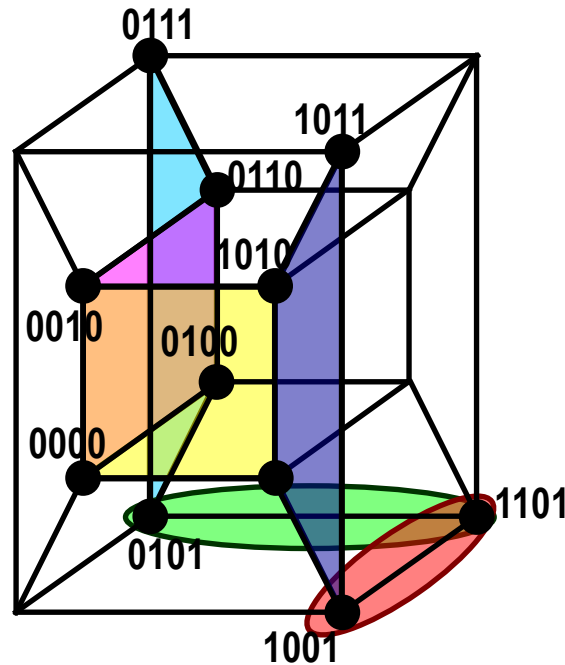
Espresso-exact

- **Exact 2-level logic minimizer**
- **Exploits iterative reduction and branch and bound algorithm on cyclic core**
- **Compact implicant table**
 - **Rows represent groups of minterms covered by the same implicants**
- **Very efficient**
 - **Solves most benchmarks**

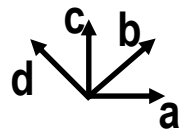
Example

After removing the essentials

	α	β	ϵ	ζ
0000,0010	1	1	0	0
1101	0	0	1	1



α	0	*	*	0	1
β	*	0	*	0	1
γ	0	1	*	*	1
δ	1	0	*	*	1
ϵ	1	*	0	1	1
ζ	*	1	0	1	1



Exact two-level minimization

- There are two main difficulties:
 - Storage of the implicant table
 - Solving the cyclic core
- Implicit representation of prime implicants
 - Methods based on binary decision diagrams
 - Avoid explicit tabulation
- Recent methods make 2-level optimization solve exactly almost all benchmarks
 - Heuristic optimization is just used to achieve solutions faster

Module 3

□ Boolean Relations

- Motivation of using relations
- Optimization of realization of Boolean relation
- Comparisons to two-level optimization

Boolean relations

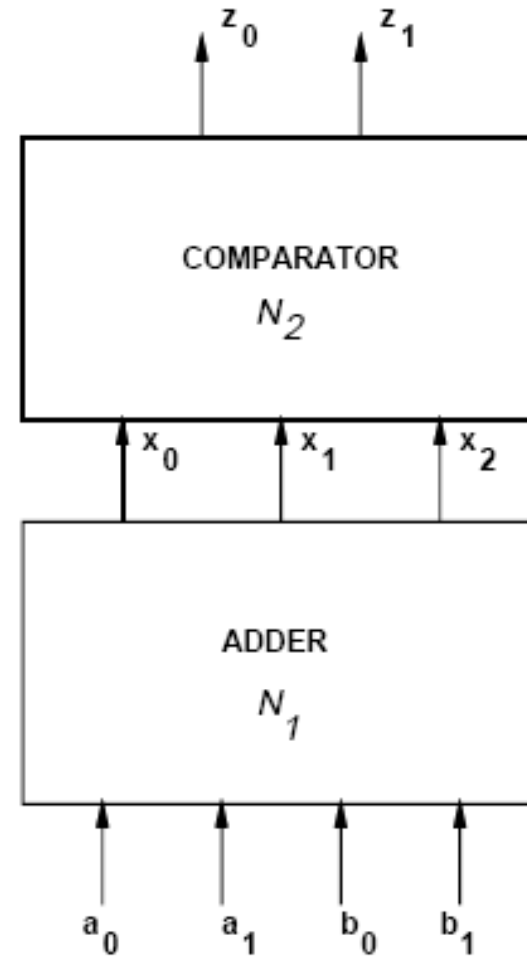
- **Generalization of Boolean functions**
- **More than one output pattern may correspond to an input pattern**
 - Multiple-choice specifications
 - Model inner blocks of multi-level circuits
- **Degrees of freedom in finding an implementation**
 - More general than *don't care* conditions
- **Problem:**
 - **Given a Boolean relation, find a minimum cover of a compatible Boolean function that can implement the relation**

Example

□ Compare:

□ $a + b > 4$?

□ $a + b < 3$?



Example

a_1	a_0	b_1	b_0	\mathbf{x}
0	0	0	0	{ 000, 001, 010 }
0	0	0	1	{ 000, 001, 010 }
0	0	1	0	{ 000, 001, 010 }
0	1	0	0	{ 000, 001, 010 }
1	0	0	0	{ 000, 001, 010 }
0	1	0	1	{ 000, 001, 010 }
0	0	1	1	{ 011, 100 }
0	1	1	0	{ 011, 100 }
1	0	0	1	{ 011, 100 }
1	0	1	0	{ 011, 100 }
1	1	0	0	{ 011, 100 }
0	1	1	1	{ 011, 100 }
1	1	0	1	{ 011, 100 }
1	0	1	1	{ 101, 110, 111 }
1	1	1	0	{ 101, 110, 111 }
1	1	1	1	{ 101, 110, 111 }

Example

□ Circuit is no longer an adder

a_1	a_0	b_1	b_0	x
0	*	1	*	010
1	*	0	*	010
1	*	1	*	100
*	*	*	1	001
*	1	*	*	001

Minimization of Boolean relations

- Since there are many possible output values (for any input), there are many logic functions implementing the relation
 - Compatible functions
- Problem
 - Find a minimum compatible function
- Do not enumerate all compatible functions
 - Compute the primes of the compatible functions
 - C-primes
 - Derive a logic cover from the c-primes

Binate covering

- **Covering problem is more complex**
 - As compared to minimizing logic functions.
- **In classic Boolean minimization we just need enough implicants to cover the minterm**
 - Covering clause is **unate** in all variables
 - Any additional implicant does not hurt
- **In Boolean relation optimization, we need to pick implicants to realize a compatible function**
 - Some implicants cannot be taken together
 - Covering clause is **binate** (implicant mutual exclusion)
 - Non-compact Boolean space

Solving binate covering

- Binate cover can be solved with branch and bound
 - In practice much more difficult to solve, because it is harder to bound effectively
- Binate cover can be reduced to min-cost SAT
 - SAT solvers can be used
- Binate cover can be also modeled by BDDs
- Several approximation algorithms for binate cover

Boolean relations

- **Generalization of Boolean functions**
 - More degrees of freedom than don't care sets
- **Useful to represent multiple choice**
- **Useful to model internals of logic networks**
- **Elegant formalism, but computationally-intensive solution method**