

#### ECE 595Z Digital VLSI Design Automation

#### Module 5 (Lectures 14-20): Multi-level Synthesis Lecture 17



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#### Lecture #16: Summary

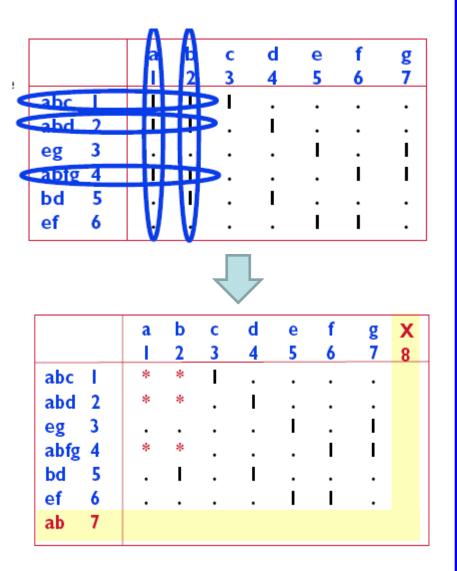
- 0-1 Matrices can be used to represent sets
  - Prime rectangles represent set intersections
- Applications to extraction
  - Co-kernels / kernels are prime rectangles in cube-literal matrix for a *single* expression
  - Single cube factors are prime rectangles in cube-literal matrix for *multiple* expressions
  - Multi-cube factors are prime rectangles in co-kernel-cube matrix
- How to select the "best" factor?
  - Value of rectangles represent change in #literals

#### Using "Value" to Select Factors

- Value is really a measure of change in #literals if we select a given factor!
- Heuristic approach to factoring:
  - 1. Find a rectangle of maximal value
  - 2. Extract it
  - 3. Update matrix
  - Repeat Steps 1-3 with the next best rectangle

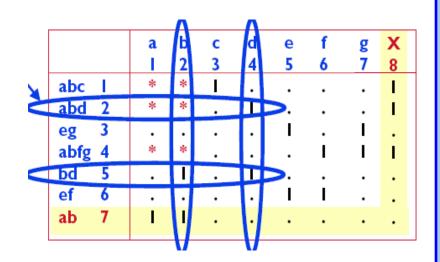
#### Matrix Update: Single-cube Case

- Once you extract a factor, need to update the cube-literal matrix, since the network has changed
  - Add a column at end of matrix, label it
  - Add another row at bottom of matrix, label it
  - Change all the "1" entries of the prime rectangle we just extracted into "\*" to indicate don't cares
    - Why? It's OK to cover these guys again with the next rectangle we extract, but if you don't cover them it's OK too



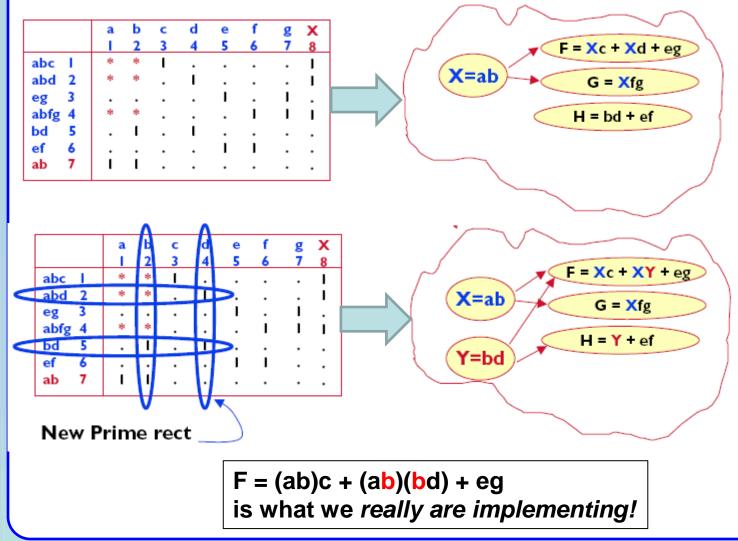
#### Side-effects of Rectangle Overlap: Single-cube Case

- It is OK for a rectangle to cover some "\*" entries
- Redefine "rectangle" to mean "rows, cols don't cover any 0 entries", *i.e.*, it's OK to cover the "\*" entries or don't cares
- However, this "messes up" an assumption of the algebraic model!



#### Side-effects of Rectangle Overlap

• Example:

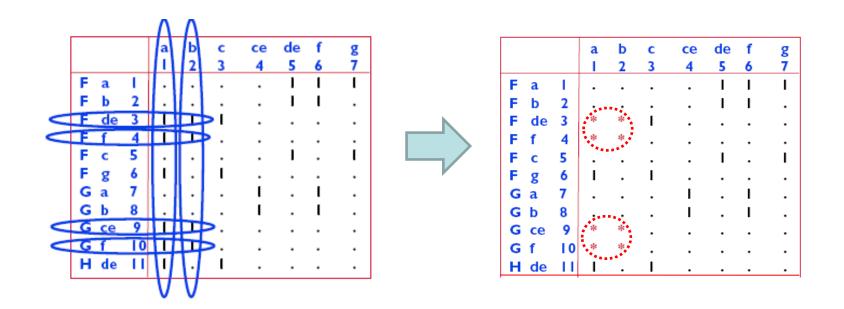


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#### Side-effects of Rectangle Overlap

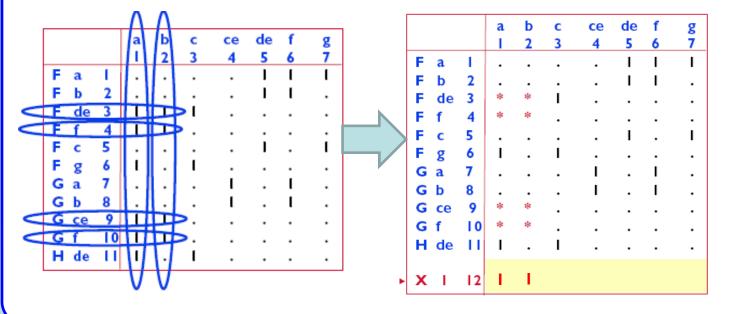
- Effect: Literals get repeated in the factoring of products
  - This is a technical violation of the algebraic model, which said that if we factor f = d.q + r, d and q should have no common variables
- Overlapping rectangles mean d, q <u>do</u> have common variables
- Bottomline: Rectangle based algorithm gives you (algebraic factoring)++
  - Actually a good thing!

- Consider co-kernel-cube matrix used for multi-cube extraction
- What do you do after extracting a factor corresponding to a prime rectangle?
  - 1. Change all the "1"s in the prime rectangle into "\*"s since they are already "covered" by the factor, they become don't cares



- What do you do after extracting a factor corresponding to a prime rectangle?
  - 1. Change all the "1"s in the prime rectangle into "\*"s
  - 2. Compute kernels for the multi-cube factor
    - For each kernel you found here, add a new row (and label as before)
    - No new columns the factor is just an intersection of kernels found before, so no new kernel cubes

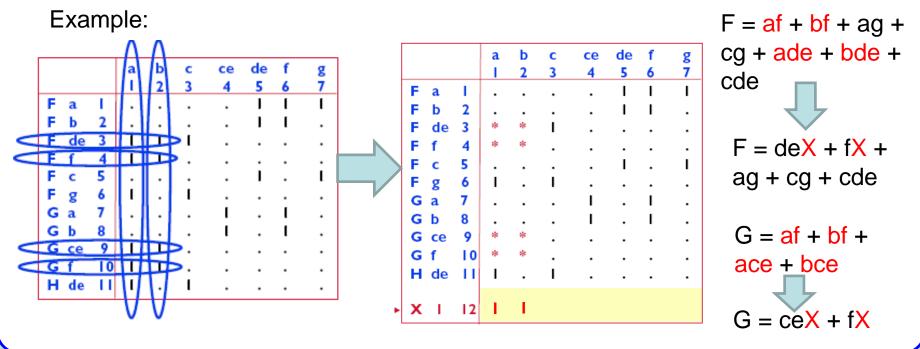




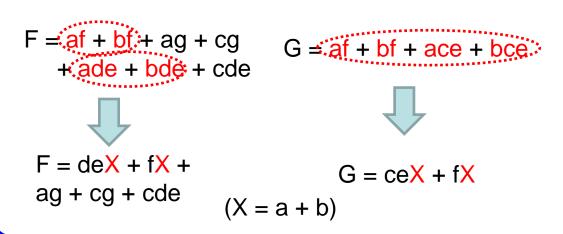
X = a + b (only 1 kernel – itself)

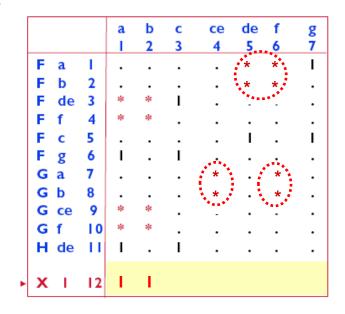
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#### 3. Other entries outside rectangle also become "\*"s (WHY?)



- Once some terms in the original SOP expressions have been "covered" by a factor, no need to consider them any more
- How can we reflect this in the cokernel-cube matrix?
- Each "1" entry in the matrix corresponds to a product term in one of the original SOP expressions



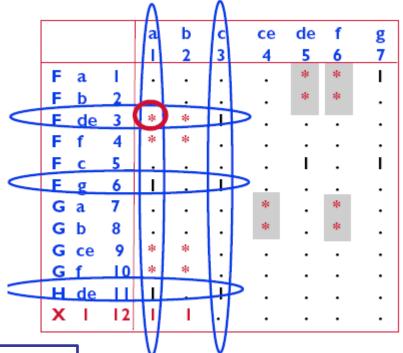


For each product term that is covered by the extracted factor, change all entries corresponding to that term to "\*"

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## Side-effects of Rectangle Overlap (Multi-cube Factors)

- As with the singlecube case, it is OK for a rectangle to include "\*" entries
- Once again, this breaks an assumption of the algebraic model



After 1<sup>st</sup> factor (X = a+b): F = deX + fX + ag + cg + cde After 2<sup>nd</sup> factor (Y = a+c): F = deX + fX + deY + gY = de(a+b) + f(a+b) + de(a+c) + g(a+c) = ade + bde + af + bf + ade + cde + ag + cg

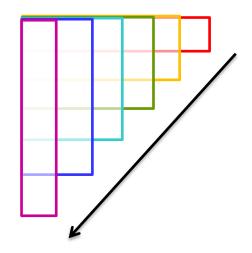
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#### **Re-cap: Extraction of Factors**

- Algebraic model
- Kernels / Co-kernels
- Unifying theme: Rectangles in 0-1 matrices
- Single-cube
  - Cube-literal matrix
  - Prime rectangle is a good factor
- Multi-cube
  - Co-kernel-cube matrix
  - Prime rectangle is a good factor
- Rectangle weights & values estimate literals saved
- Overall approach: Iteratively generate "good" rectangles (another covering problem!)

#### Algorithm for Rectangle Generation : Greedy Row

- Objective: Find prime rectangles one at a time
  - Start with a "seed" row (rectangle with 1 row and highest value)
  - 2. Add a row such that resulting rectangle is of highest value (over all possible row additions).
  - 3. Repeat until rectangle is a single column
  - 4. Select best rectangle seen



**Dual algorithm -Greedy Column**: Start with seed column, grow by adding columns

### Algorithm for Rectangle Generation : Ping-Pong

- Objective: Find prime rectangles one at a time
  - 1. Start with a "seed" row (rectangle with 1 row and highest value)
  - 2. Add a row such that resulting rectangle is of highest value (over all possible row additions). Keep adding until value increases.
  - 3. Add a column such that resulting rectangle is of highest value. Keep adding until value increases.
  - 4. Go to step 2.
  - 5. Stop when you rectangle cannot be "profitably" grown in any direction.

Used in MIS & SIS

## **Boolean Optimization**

#### Can we go beyond the Algebraic Model?

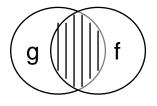
- The algebraic model enables efficient (**fast**) transformations of a Boolean network
  - Collapsing/elimination, extraction/decomposition, substitution, ...
- However, it is limited in the **scope of optimization** since it does not take advantage of the unique properties of Boolean algebra.

#### **Boolean Division**

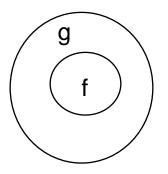
 p is a Boolean divisor of f if q ≠ 0 and r exist such that

f = pq + r

- q and r are not unique
- If r = 0, q is an exact factor of f
- Theorem: If f.g ≠ 0, then g is a Boolean divisor of f
- Theorem: A logic function g is an exact Boolean factor of a logic function of f iff f ⊆ g
- Boolean division with a given factor is OK, but no efficient method for identifying Boolean factors is known



f = gq + r



Number of Boolean divisors and factors is LARGE!

#### **Boolean Division: An indirect approach**

- Apply two-level minimization!
- Given f = (f<sup>ON</sup>, f<sup>DC</sup>, f<sup>OFF</sup>), and g, find the "best" h,r such that gh + r is a cover for f.
  - Create new variable y to represent output of g
  - Add  $y \neq g$  to the don't care set for f
    - $f^{DC*} = yg' + y'g$
  - Minimize  $(f^{ON}(f^{DC^*})', f^{DC} + f^{DC^*}, f^{OFF}(f^{DC^*})')$ 
    - Use ESPRESSO (modify cost-function to literals)
    - Force y to appear in the result
    - Product terms with y form the quotient, other terms form the remainder



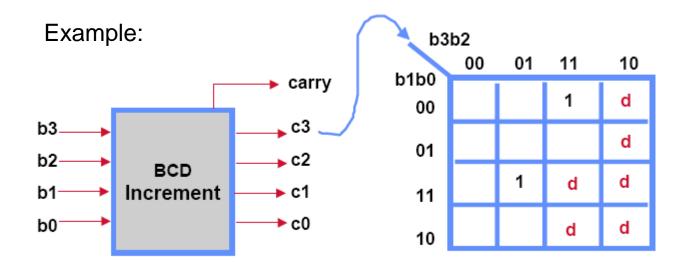
#### Boolean Optimization Using Don't Cares

#### Augmenting Algebraic Methods with Boolean Methods

- General strategy for technologyindependent optimization
  - Circuit re-structuring using algebraic methods
  - Circuit simplification using Boolean methods

#### Don't Cares

• From undergraduate digital logic design: Don't cares are input patterns that could "never happen"



Patterns b3 b2 b1 b0 = 1010, 1011, 1100, 1101, 1110, 1111 cannot happen

#### Don't Cares in Multi-level Boolean Networks

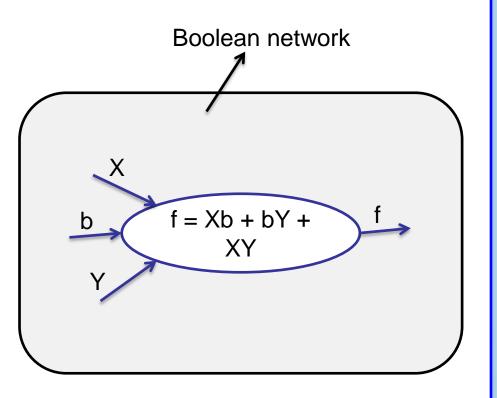
- The "usual" don't cares
  - Somebody tells you based on the semantics of the primary inputs
  - **Explicit** don't cares
- Don't cares that occur due to the **network structure** 
  - What we will talk about in this class
  - Implicit don't cares

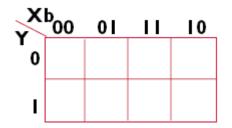
## Outline

- Informal introduction to implicit don't cares in Boolean networks
- A more rigorous definition
- Optimizing Boolean networks using implicit don't cares
- Prime and irredundant Boolean networks

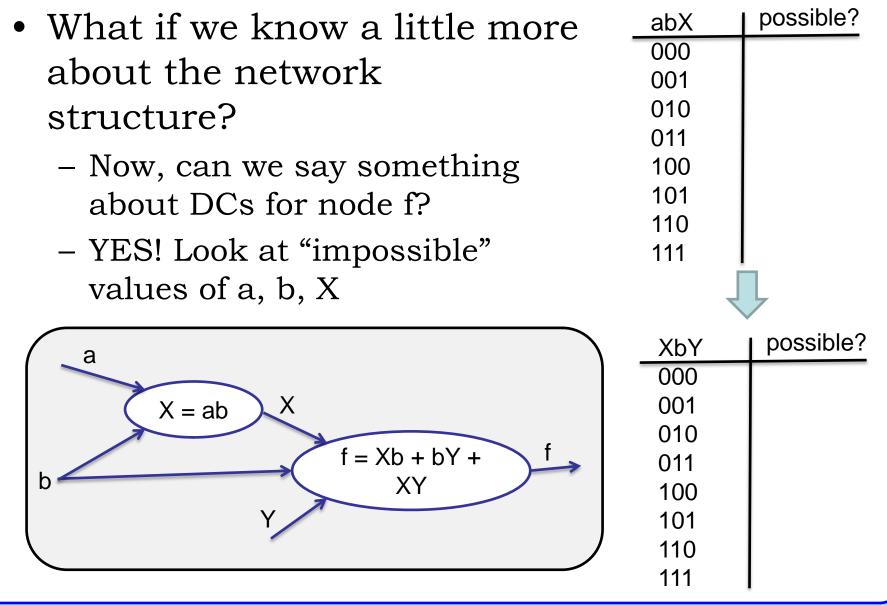
#### Informal Introduction to DCs

- Given a node in a Boolean network
- Can we say anything about don't cares for node f?



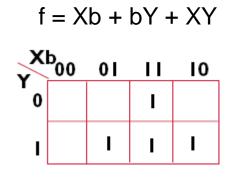


#### Informal Introduction to DCs

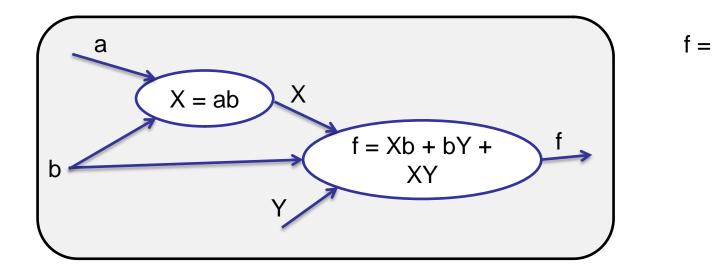


#### Informal Introduction to DCs

 How can we simplify f using these don't cares?



Add don't cares & simplify



#### Lecture #17: Summary

#### • Finished the discussion on Algebraic optimization

- Ping-pong algorithm for rectangle generation
- Boolean Optimization
  - Boolean division and factoring are much harder than algebraic counterparts
  - Number of Boolean divisors/factors can be very large
  - Boolean division using 2-level minimization
- Boolean Optimization Using Implicit Don't Cares
  - Implicit don't cares are introduced due to the network itself
  - We saw how to derive don't cares at a node's inputs by looking at it's fanin nodes
  - A node can be simplified using its implicit don't cares