

# Reading/Hand-on Assignment 1

- Survey of 3 SAT solvers
  - **MiniSAT**, Sweden.
  - **CHAFF**, Princeton University.
  - **GRASP**, University of Michigan.
- 3 groups, 1 group per solver.
- Oral presentation (**April 14<sup>th</sup>**, in class)
  - Technical details.
  - Your test run of the solvers + results.
- Written report (due **April 19<sup>th</sup>**)
  - One copy per group.

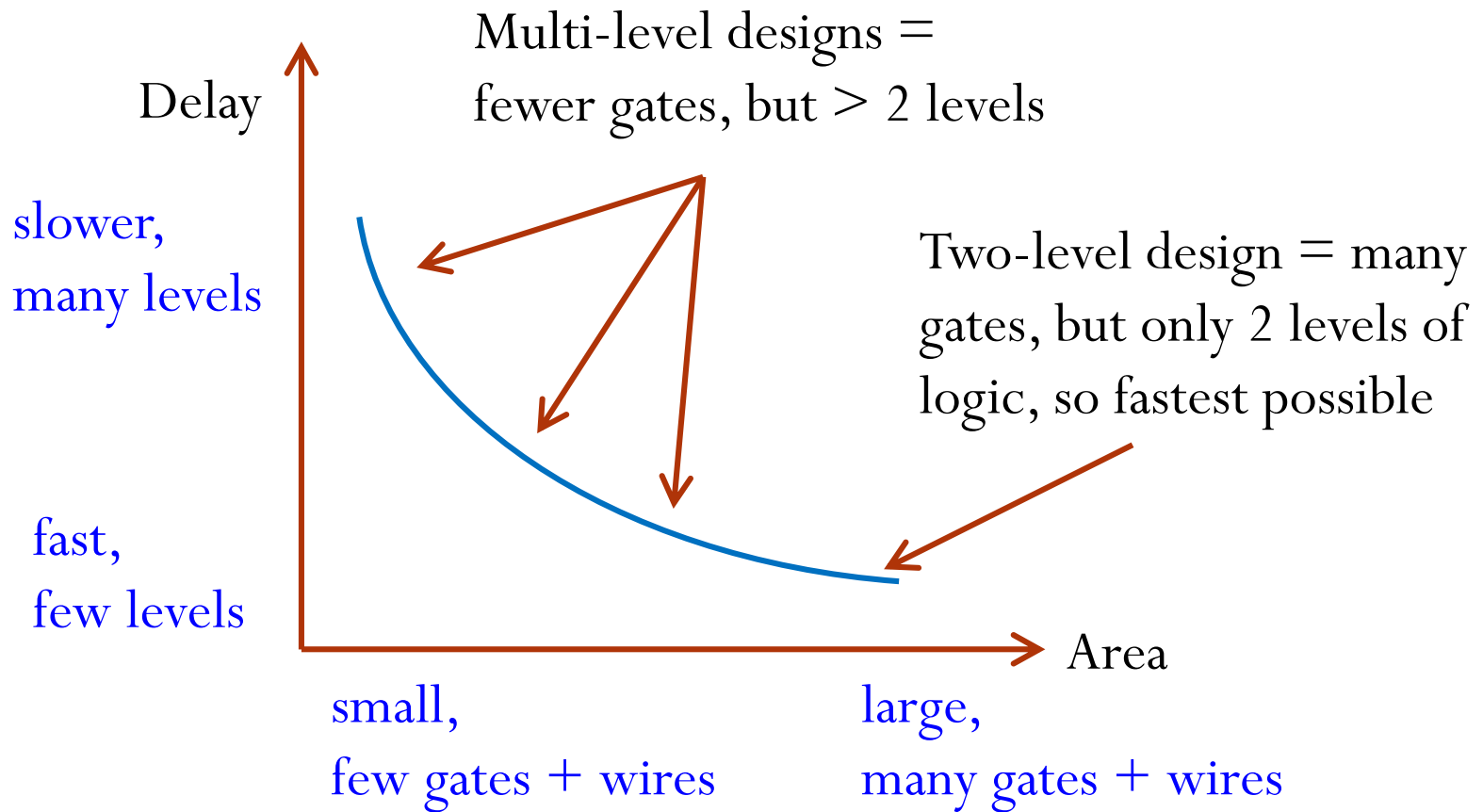
# Multi-Level Logic Synthesis

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# Why Multi-level Logic?

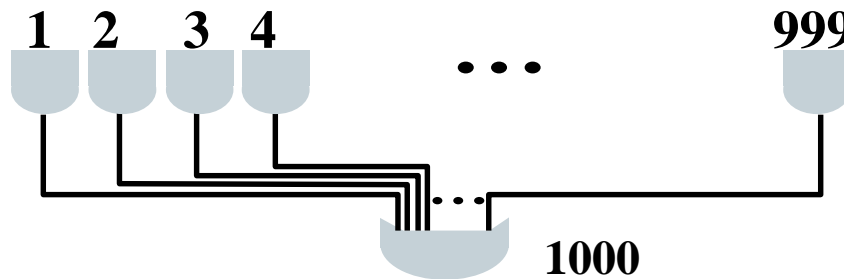
- Two-level forms are too **restrictive**.
- It has small delay but large area.
  - **Area** = gates + literals (wires), i.e., things that take space on a chip.
  - **Delay** = maximum levels of logic gates required to compute function.
  - Two-level is **minimum** gate delay possible, but usually **worst** on area.

# Area versus Delay Tradeoff



# Why Multi-level Logic?

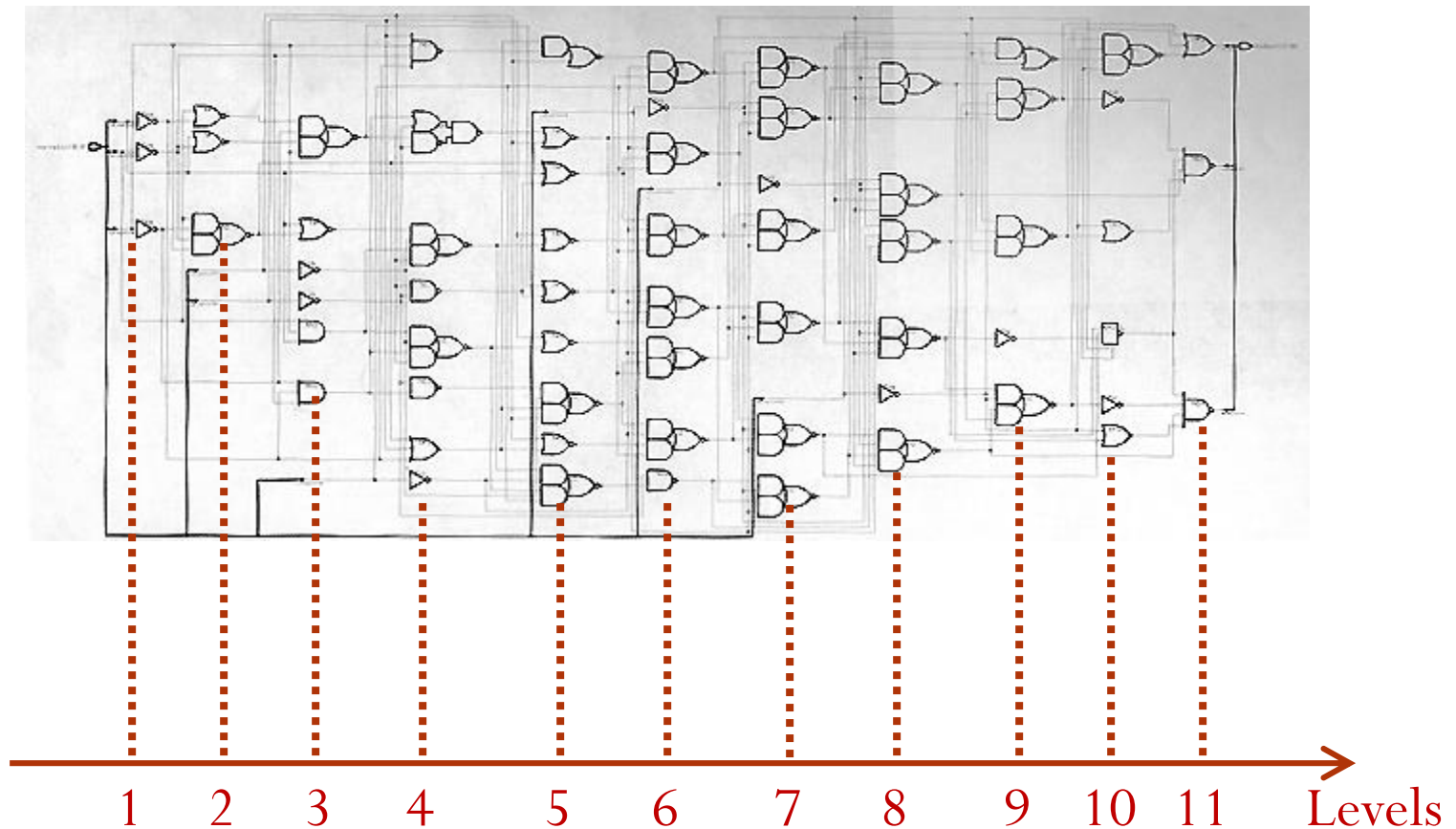
- **Rarely** see 2-level designs for really big things...
  - We use 2-level logic mostly for **pieces** of bigger things.
  - Even small things routinely done as **multi-level**.
- What does a 2-level design with 1000 gates look like?



This is just **NOT** going to be the preferred logic network structure...

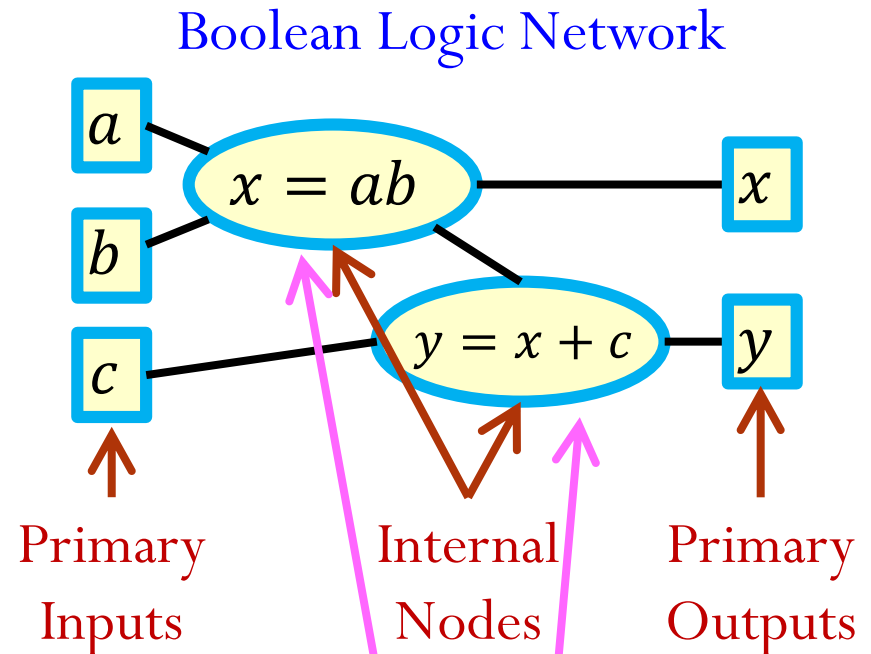
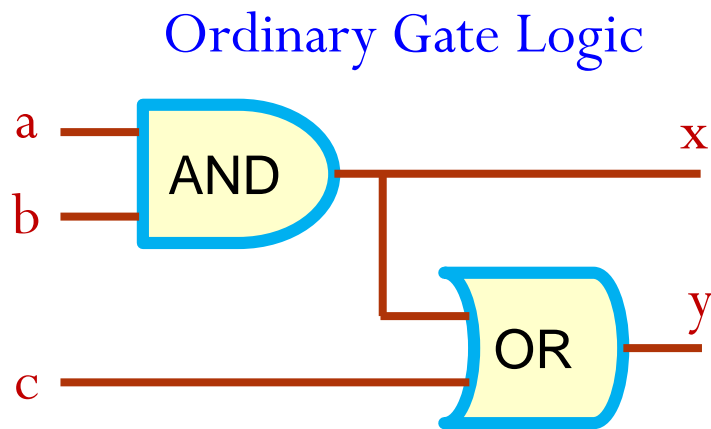
# Real Multilevel Example

- A small design, done by commercial synthesis tool.



# Boolean Logic Network Model

- Need more sophisticated model: **Boolean Logic Network**
  - Idea: it's a **graph** of connected blocks, like any logic diagram, but now individual component blocks can be **2-level Boolean functions in SOP form**.



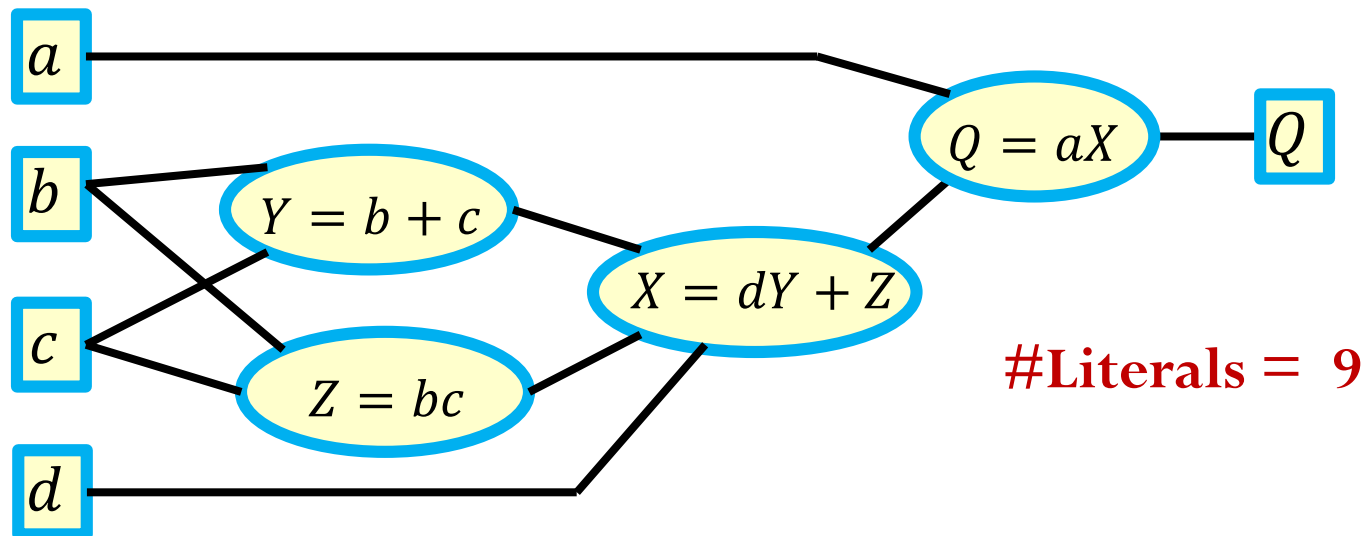
$x$  and  $y$  are now Boolean function.

# Multilevel Logic: What to Optimize?

- A simplistic but surprisingly useful metric:

## Total literal count

- Count **every** appearance of **every** variable on right hand side of “=” in **every internal node**.
- Delays also matter, but for this class, only focus on **logic complexity**.





# Optimizing Multilevel Logic: Big Ideas

- Again: **Boolean logic network** is a **data structure**. What **operators** do we need?
- 3 basic kinds of operators:
  - **Simplify** network nodes: no change in # of nodes, just simplify insides, which are **SOP form**.
  - **Remove** network nodes: take “too small” nodes, **substitute them back** into fanouts.
    - This is not too hard. This is mostly manipulating the graph, simple SOP edits.
  - **Add** new network nodes: this is **factoring**. Take big nodes, split into smaller nodes.
    - This is a **big deal**. This is new. This is what we need to teach you...

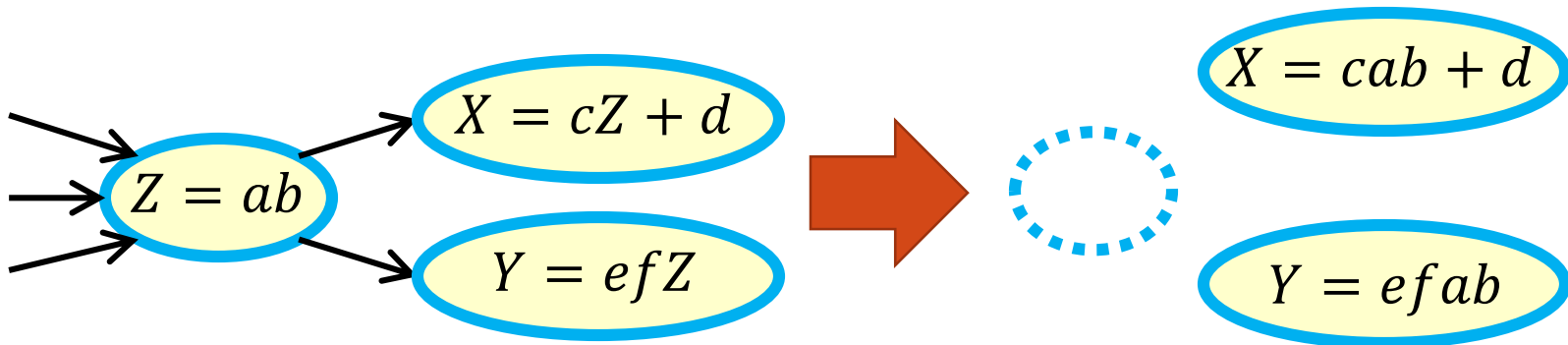
# Simplifying a Node

- You already know this! This is **2-level synthesis**.
- Just run ESPRESSO on 2-level form **inside** the node, to reduce # literals.
- As structural changes happen across network, “**insides**” of nodes may present opportunity to simplify.



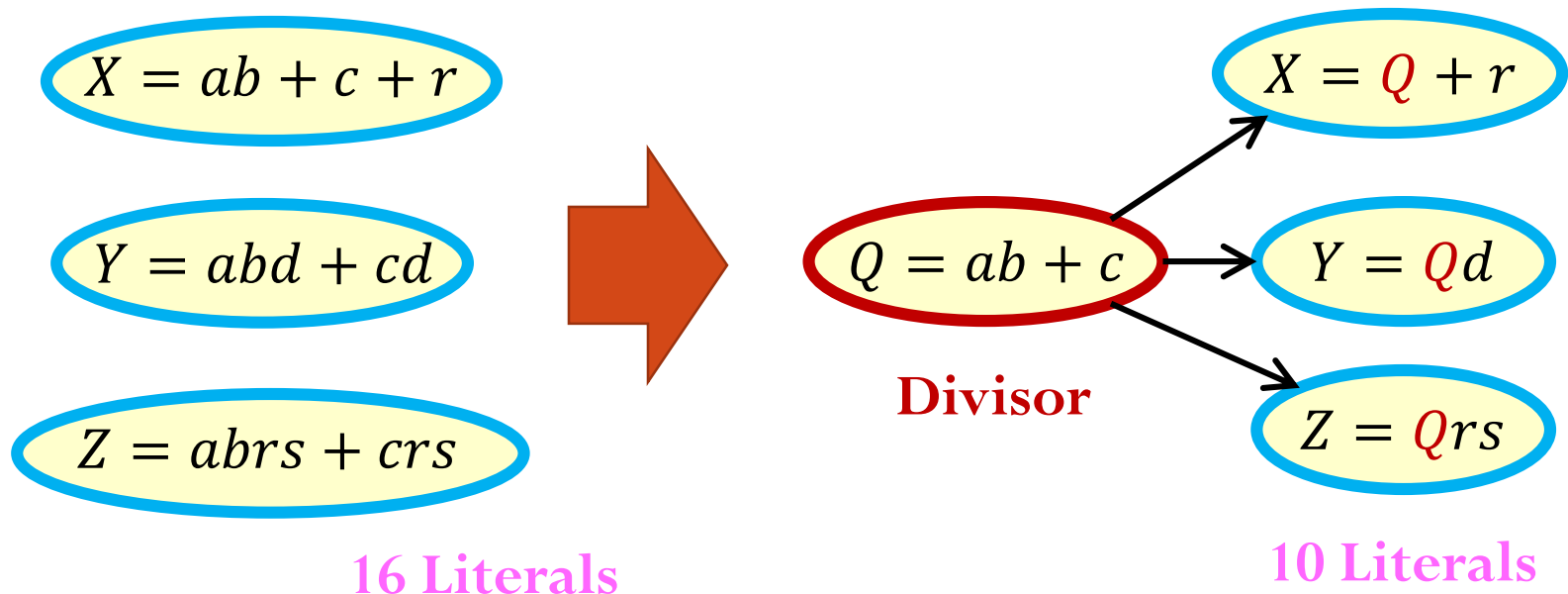
# Removing a Node

- Typical case is you have a “**small**” factor which doesn’t seem to be worth making it a **separate** node.
- “**Push**” it back into its fanouts, make those nodes bigger, and hope you can use 2-level simplification on them.



# Adding new Nodes

- This is Factoring, this is new, and hard.
  - Look at existing nodes, identify **common divisors**, extract them, connect as **fan-ins**.
  - Tradeoff: **more** delay (another level of logic), but **fewer literals** (less gate area).



# Multi-Level Logic Synthesis

- A more common design style.
  - Small area, but may have large delay.
- More sophisticated model: **Boolean logic network**
- 3 kinds of optimizing step:
  - Simplify a node by **2-level minimization**.
  - Remove a node by **substituting**.
  - Add a node by **factoring**.

# Multilevel Synthesis Scripts

- Multilevel synthesis like 2-level synthesis is **heuristic**.
- ...but it's also more complex. Write **scripts of basic operators**.
  - Do **several passes** of different optimizations over the Boolean logic network.
  - Do some “cleanup” steps to get rid of “too small” nodes (**remove nodes**).
  - Look for “easy” **factors**: just look at existing nodes, and try to use them.
  - Look for “hard” **factors**: do some work to extract them, try them, and keep the good ones.
  - Do 2-level optimization of insides of each logic node in network (**simplify nodes** by ESPRESSO).
  - Lots of “art” in the engineering of these scripts.
- For us, the one big thing you don't know: **How to factor...**

# Multi-Level Logic Synthesis

- We need a new operator: **factoring**
- Problem #1: how to do division?
  - **Solution**: Algebraic model and algebraic division
  - **Algebraic model**: **Pretending** that Boolean expressions behave like **polynomials of real numbers**, not like Boolean algebra.
  - **Algebraic division**: Given a Boolean expression  $F$  and a divisor  $D$ , obtain quotient  $Q$  and remainder  $R$ , such that
$$F = D \cdot Q + R$$
- Problem #2: how to find good divisors?
  - **Solution**: **Kernels**.

# Another New Model: Algebraic Model

- Factoring: How do we really do it?
  - Develop another model for Boolean functions, cleverly designed to let us do this
  - Tradeoff: lose some “expressivity” – some aspects of Boolean behavior and some Boolean optimizations we just **cannot do**, but we **gain practical factoring**.
- New model: **Algebraic model**
  - Term “algebraic” comes from pretending that Boolean expressions behave like **polynomials of real numbers**, not like Boolean algebra.
  - **Big new Boolean operator: Algebraic Division** (or, also “**Weak**” Division).

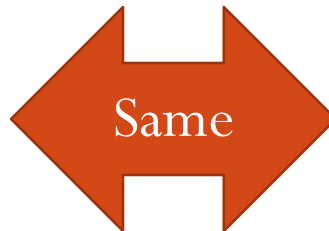


# Algebraic Model

- Idea: keep just those **rules** that work for **BOTH** polynomials of reals **AND** Boolean algebra, but **get rid of** the rest.

## Real numbers

$$\begin{aligned}a \cdot b &= b \cdot a & a + b &= b + a \\a \cdot (b \cdot c) &= (a \cdot b) \cdot c \\a + (b + c) &= (a + b) + c \\a \cdot (b + c) &= a \cdot b + a \cdot c \\a \cdot 1 &= a & a \cdot 0 &= 0 \\a + 0 &= a\end{aligned}$$



## Boolean algebra

$$\begin{aligned}a \cdot b &= b \cdot a & a + b &= b + a \\a \cdot (b \cdot c) &= (a \cdot b) \cdot c \\a + (b + c) &= (a + b) + c \\a \cdot (b + c) &= a \cdot b + a \cdot c \\a \cdot 1 &= a & a \cdot 0 &= 0 \\a + 0 &= a\end{aligned}$$

$$a \cdot \bar{a} = 0 \quad a + \bar{a} = 1$$

$$a \cdot a = a \quad a + a = a$$

$$a + 1 = 1$$

$$(a + b)(a + c) = a + b \cdot c$$



# Algebraic Model

- If we only get to use algebra rules from real numbers...
  - **Consequence:** A variable and its complement must be treated as **totally unrelated**.
    - Since no expression like  $a + \bar{a} = 1$  allowed.

$$F = ab + \bar{a}x + \bar{b}y$$



$$\text{Let } R = \bar{a}, S = \bar{b}$$

$$F = ab + Rx + Sy$$

- **Aside:** this is one of the losses of “expressive power” of Boolean algebra.

# Algebraic Model

- Idea
  - Boolean functions manipulated in SOP form like **polynomials**.
  - Each product term in such an expression is just **a set of variables**, e.g.,  $abRy$  is the set  $(a, b, R, y)$ .
  - SOP expression itself is just a **list of these products (cubes)**, e.g.,  $ab + Rx$  is the list  $(ab, Rx)$ .

# Algebraic Division: Our Model for Factoring

- Given function  $F$  we want to factor as:

$$F = D \cdot Q + R$$

divisor      quotient      remainder

- If remainder  $R = 0$ , we call the divisor as a “**factor**”.

**Example:**  $F = ac + ad + bc + bd + e$


$$= (a + b)(c + d) + e$$

divisor      quotient      remainder

# Algebraic Division

- Example:  $F = ac + ad + bc + bd + e$ 
  - Want:  $F = D \cdot Q + R$ .

Divisor is a **factor** if  $R = 0$ .



Divisor ( $D$ )	Quotient ( $Q$ )	Remainder ( $R$ )	Is $D$ Factor?
$ac + ad + bc + bd + e$	1	0	Yes
$a + b$	$c + d$	$e$	No
$c + d$	$a + b$	$e$	No
$a$	$c + d$	$bc + bd + e$	No
$b$	$c + d$	$ac + ad + e$	No
$c$	$a + b$	$ad + bd + e$	No
$d$	$a + b$	$ac + bc + e$	No
$e$	1	$ac + ad + bc + bd$	No

# Algebraic Division: Very Nice Algorithm

- **Inputs**: A Boolean expression  $F$  and a divisor  $D$ , represented as lists of cubes (and each cube as a set of literals).
- **Output**
  - Quotient  $Q = F/D$  as a cube list, or **empty** if  $Q = 0$ .
  - Remainder  $R$  as a cube list, or **empty** if  $D$  was a **factor**.
- **Strategy**
  - Cube-wise walk through cubes in divisor  $D$ , trying to divide each of them into  $F$ .
  - ... intersect all the division results.

# Algebraic Division Algorithm

```
AlgebraicDivision( F, D ) { // divide D into F
  for ( each cube d in divisor D ) {
    let C = { cubes in F that contain this product term d };
    if ( C is empty ) return ( quotient = 0, remainder = F );
    let C = cross out literals of cube d in each cube of C;
    if ( d is the first cube we have looked at in divisor D )
      let Q = C;
    else Q = Q  $\cap$  C;
  }
  R = F - ( Q  $\cdot$  D );
  return ( quotient = Q, remainder = R );
}
```


## Example:

Cube **xyzw** contains product term **yz**



## Example:

Suppose **C = xyz + yzw + pqyz** and **d = yz**. Then crossing out all the **yz** parts yields **x + w + pq**



# Algebraic Division: Example

$$F = axc + axd + axe + bc + bd + de, D = ax + b$$

$F$ cube	$D$ cube: $ax$	$D$ cube: $b$
$axc$	$axc \rightarrow c$	—
$axd$	$axd \rightarrow d$	—
$axe$	$axe \rightarrow e$	—
$bc$	—	$bc \rightarrow c$
$bd$	—	$bd \rightarrow d$
$de$	—	—

$$C = c + d + e \quad C = c + d$$

$$Q = (c + d + e) \cap (c + d) = c + d$$

$$R = F - QD = axe + de$$



# Algebraic Division: Warning

- Remember: No “**Boolean**” simplification, only “**algebraic**”.

- So what? Well, suppose you have this

$$F = a\bar{b}\bar{c} + ab + ac + bc, D = ab + \bar{c}$$

and you want  $F/D$ .

- You must let  $X = \bar{b}$  and  $Y = \bar{c}$  and transform  $F$  and  $D$  to something like

$$F = aXY + ab + ac + bc, D = ab + Y$$

- Because we must treat the true and complement forms of variables as **totally unrelated**.

# One More Constraint: Redundant Cubes

- To do  $F/D$ , function  $F$  must have no **redundant** cubes
  - Technical definition is: **minimal** with respect to **single-cube containment**.
  - Means: no one cube is **completely** covered by **one of the other** cubes in SOP cover.
    - E.g.,  $abcd$  is completely covered by  $ab$ .
- Why **no** redundant cubes?
  - Consider:  $F = a + ab + bc$  and  $D = a$ .
    - Note:  $F$  has redundant cube  $ab$ .
  - What is  $F/D$  by our **algebraic division algorithm**?

$$Q = F/D = 1 + b$$

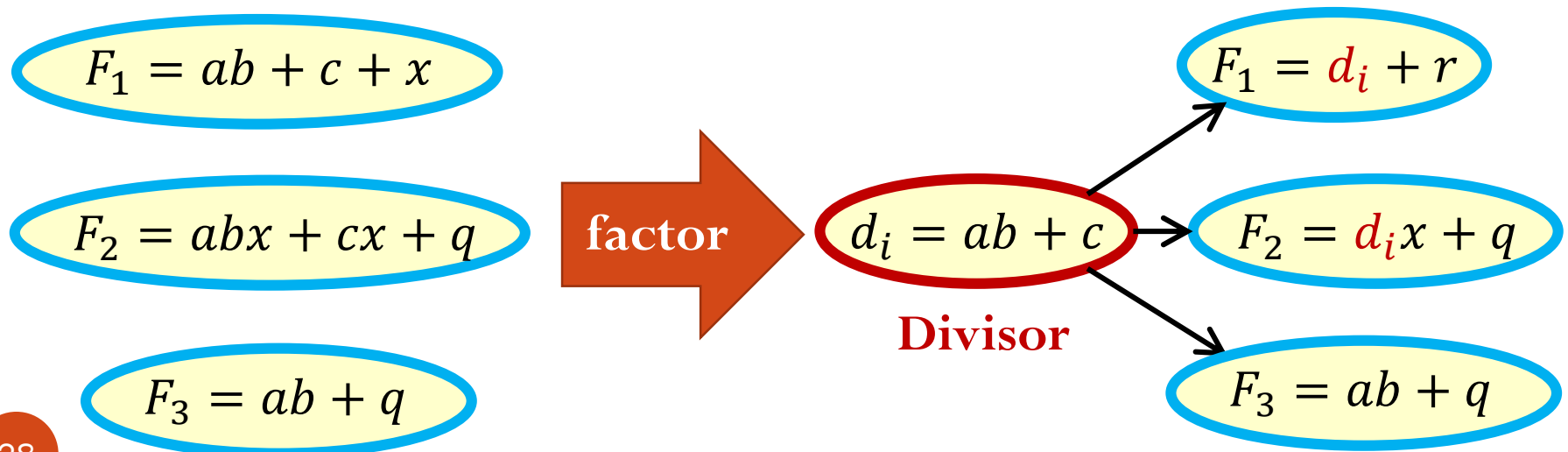
However, we don't have  $1 + \text{stuff}$  operation in algebraic model!

# One More Constraint: Redundant Cubes

- ... So, we should remove redundant cubes to make the SOP **minimal** with respect to **single-cube containment**.
  - Not hard.

# Multilevel Logic Synthesis: Where are We?

- For Boolean function  $F$  and  $D$ , can compute  $F = Q \cdot D + R$  via **algebraic model**.
  - It is great, but still **not enough**: don't know how to **find** a good divisor  $D$ .
  - **Another problem**: given  $n$  functions  $F_1, F_2, \dots, F_n$ , find **a set of good common divisors**.



# Where to Look for Good Divisors?

- Surprisingly, the **algebraic model** has a beautiful answer.
  - One more reason we like it: Has some surprising and elegant “deep structure”.
- Where to look for divisors of function  $F$ ?
  - In the set of **kernels** of  $F$ , denoted  $K(F)$ .
  - $K(F)$  is another set of 2-level SOP forms which are the special, foundational structure of any function  $F$ , being interpreted in our algebraic model.
- How to find a kernel  $k \in K(F)$ ?
  - **Algebraically divide**  $F$  by one of its **co-kernels**,  $c$ .

# Kernels and Co-Kernels of Function $F$

- **Kernel** of a Boolean expression  $F$  is:
  - A **cube-free** quotient  $k$  obtained by **algebraically dividing**  $F$  by a **single cube**  $c$ .
  - This single cube  $c$  also has a name: it is a **co-kernel** of function  $F$ .

$$\begin{array}{r}
 \text{quotient } Q \\
 \hline
 \text{divisor } D \left[ \begin{array}{c} \text{expression } F \\ \vdots \end{array} \right. \\
 \hline
 \text{remainder } R \\
 F = D \cdot Q + R
 \end{array}$$

$$\begin{array}{r}
 \text{kernel } k \text{ (cube-free)} \\
 \hline
 c = 1 \text{ cube} \left[ \begin{array}{c} \text{expression } F \\ \vdots \end{array} \right. \\
 \hline
 \text{remainder } R \\
 F = c \cdot k + R
 \end{array}$$

# Kernels Are Cube-Free...

- **Cube-free** means...?
  - You cannot factor out a **single** cube divisor that **leaves no remainder**.
  - Technically: has no **one cube** that is a **factor** of expression.
    - Pick a cube  $C$ . If you can “**cross out**”  $C$  in **each** product term of  $F$ , then  $F$  is **not** a **kernel**.

Expression $F$	$F = D \cdot Q + R$	$F$ Cube-free?
$a$	$a \cdot 1 + 0$	No
$a + b$	--	Yes
$ab + ac$	$a \cdot (b + c) + 0$	No
$abc + abd$	$ab \cdot (c + d) + 0$	No
$ab + acd + bd$	--	Yes

# Some Kernel Examples

- Suppose  $F = abc + abd + bcd$

Divisor cube $d$	$Q = F/d$	Is $Q$ a kernel of $F$ ?
1	$abc + abd + bcd$	No, has cube = $b$ as factor
$a$	$bc + bd$	No, has cube = $b$ as factor
$b$	$ac + ad + cd$	Yes! co-kernel = $b$
$ab$	$c + d$	Yes! co-kernel = $ab$

- Any Boolean function  $F$  can have many different kernels.
  - The set of kernels of  $F$  is denoted as  $K(F)$ .



# Kernels: Why Are They Important?

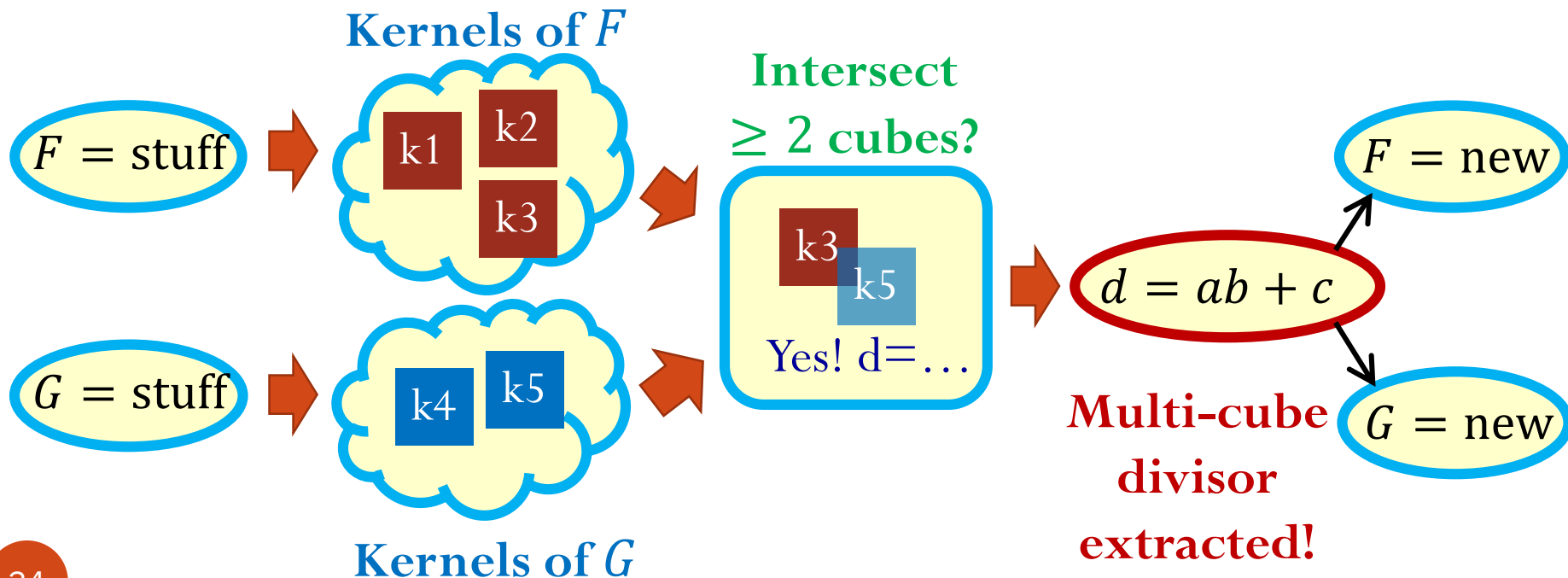
- Big result: Brayton & McMullen Theorem
  - **From:** R. Brayton and C. McMullen, “The decomposition and factorization of Boolean expressions.” In *IEEE International Symposium on Circuits and Systems*, pages 49–54, 1982.

Expressions  $F$  and  $G$  have a **common multiple-cube divisor**  $d$  **if and only if:**

there are kernels  $k_1 \in K(F)$  and  $k_2 \in K(G)$  such that  $d = k_1 \cap k_2$  and  $d$  is an expression with **at least 2** cubes in it (i.e.,  $k_1$  and  $k_2$  have **common cubes**).

# Multiple-Cube Divisors and Kernels

- Brayton & McMullen Theorem in words:
  - The only place to look for **multiple-cube divisors** is in the **intersection of kernels**!
    - Indeed, this intersection of kernels contains all divisors.



# Brayton-McMullen: Informal Illustration

$$F = \text{cube1} \cdot \mathbf{\text{kernel1}} + \text{remainder1}$$

$$G = \text{cube2} \cdot \mathbf{\text{kernel2}} + \text{remainder2}$$

Assume:

$$\mathbf{\text{kernel1}} \cap \mathbf{\text{kernel2}} = \mathbf{X + Y}$$



$$F = \text{cube1} \cdot [\mathbf{X + Y + \text{stuff1}}] + \text{remainder1}$$

$$G = \text{cube2} \cdot [\mathbf{X + Y + \text{stuff2}}] + \text{remainder2}$$



$$F = \text{cube1} \cdot [\mathbf{X + Y}] + [\text{cube1} \cdot \text{stuff1} + \text{remainder1}]$$

$$G = \text{cube2} \cdot [\mathbf{X + Y}] + [\text{cube2} \cdot \text{stuff2} + \text{remainder2}]$$



$$\mathbf{d = X + Y}$$

$$F = \text{cube1} \cdot \mathbf{d} + \dots$$

$$G = \text{cube2} \cdot \mathbf{d} + \dots$$

# Kernels: Real Example

$$F = ae + be + cde + ab$$

$$G = ad + ae + bd + be + bc$$

Kernels	Co-kernel
$a + b + cd$	$e$
$b + e$	$a$
$a + e$	$b$
$ae + be + cde + ab$	$1$

Kernels	Co-kernel
$a + b$	$d$ or $e$
$d + e$	$a$
$c + d + e$	$b$
$ab + ae + bd + be + bc$	$1$

Intersecting these 2 kernels:  $(a + b + cd) \cap (a + b) = a + b$

# Kernels: Very Useful, But How To Find?

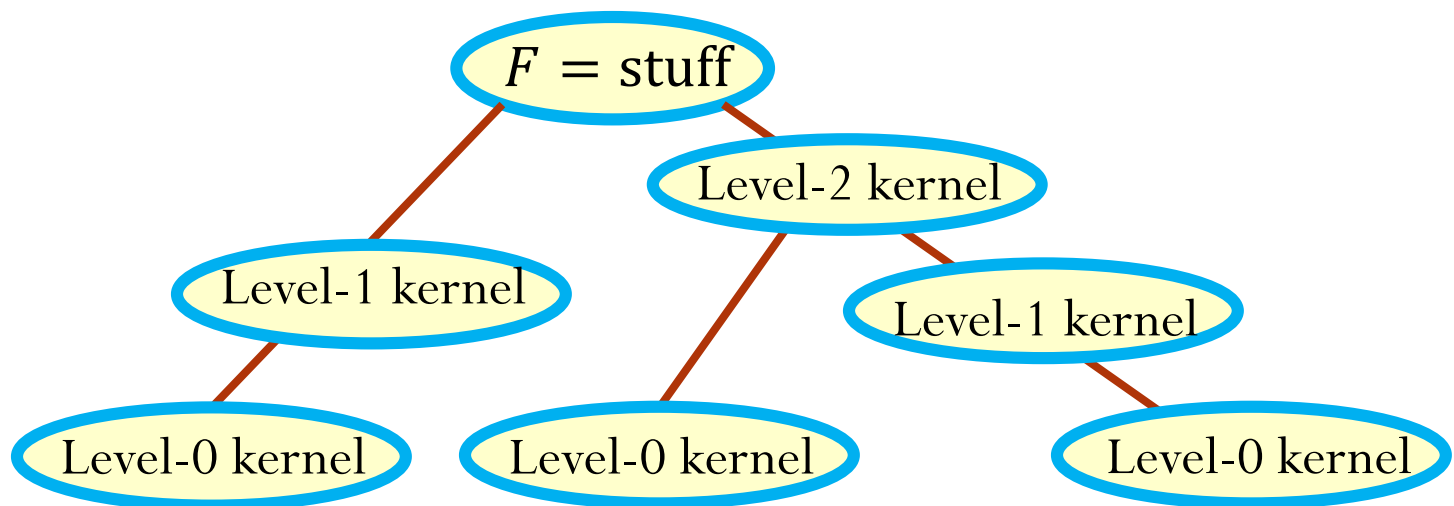
- Another **recursive** algorithm (“recursive” again!)
  - There are 2 more useful properties of kernels we need to see first...
- Start with a function  $F$  and a kernel  $k_1$  in  $K(F)$ 
$$F = \text{cube}_1 \cdot k_1 + \text{remainder}_1$$
- Then: a new, interesting question: what about  $K(k_1)$ ?
  - $k_1$  is a perfectly nice Boolean expression, so it has got **its own** kernels.
  - Do these  $k_1$ 's kernels have anything interesting to say about  $K(F)$ ?

# How $K(k_1)$ Relates to $K(F)$ ...

- We know this:  $F = \text{cube}_1 \cdot k_1 + \text{remainder}_1$
- Suppose  $k_2$  is a kernel in  $K(k_1)$ , then we know
$$k_1 = \text{cube}_2 \cdot k_2 + \text{remainder}_2$$
- Substitute this expression for  $k_1$  in original expression for  $F$ 
$$F = \text{cube}_1 \cdot [\text{cube}_2 \cdot k_2 + \text{remainder}_2] + \text{remainder}_1$$
- Since  $\text{cube}_1 \cdot \text{cube}_2$  is itself just another **single cube**, we have:
$$F = (\text{cube}_1 \cdot \text{cube}_2) \cdot [k_2] + [\text{cube}_1 \cdot \text{remainder}_2 + \text{remainder}_1]$$
- **Conclusion**:  $k_2$  also a **kernel** of original  $F$  (with **co-kernel**  $\text{cube}_1 \cdot \text{cube}_2$ )

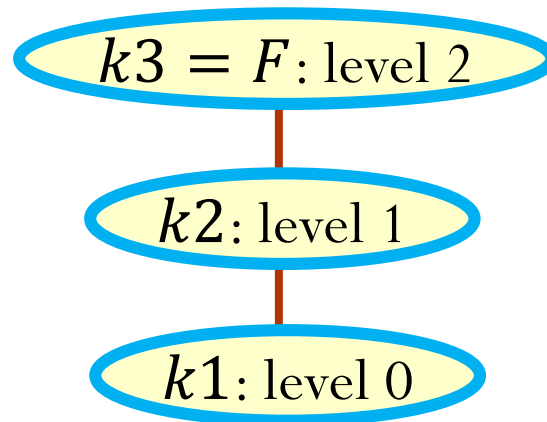
# There is a Hierarchy of Kernels Inside F

- **Definition:**  $k \in K(F)$  is
  - A **level-0 kernel** if it contains no kernels inside it except itself.
    - In words: Only cube you can pull out, get a cube-free quotient is “1”.
  - A **level-n kernel** if it contains **at least one** level-(n-1) kernel, and no other level-n kernels except itself.
    - In words: a level-1 kernel only has level-0 kernels inside it. A level-2 kernel only has level-1 and level-0 kernels in it, etc...



# Kernel Hierarchy: Example

- $F = abe + ace + de + gh$  has three kernels:
  - $k1 = b + c$ , with co-kernel  $ae$ .
  - $k2 = ab + ac + d$ , with co-kernel  $e$ .
  - $k3 = F$  with co-kernel 1.
- Note:  $k1$  is level 0,  $k2$  is level 1, and  $k3$  is level 2.





# Kernels

- Second useful result [by Brayton et al.]:
  - **Co-kernels** of a Boolean expression in SOP form correspond to **intersections of 2 or more cubes** in this SOP form.
- **Note: Intersections** here means that we regard a cube as **a set of literals**, and look at common subsets of literals.
  - This is not like “AND” for products. This just extracts **common literals**.
  - Example:  $ace + bce + de + g$

$$ace \cap bce = ce \quad \rightarrow \quad ce \text{ is a potential co-kernel}$$

$$ace \cap bce \cap de = e \quad \rightarrow \quad e \text{ is a potential co-kernel}$$

# How to Find Kernels Using These 2 Results?

- Find the kernels **recursively**.
  - Whenever find one kernel, call **FindKernels()** on it, to find (if any) lower level kernels inside.
- Use **algebraic division** to divide function by potential co-kernels, to drive recursion.
  - Use 2<sup>nd</sup> result – co-kernels are **intersections** of the cubes: If there're at least 2 cubes, then look at the intersection of those cubes, and use that intersected result as our potential co-kernel cube.
- One technical point: need to start with a **cube-free function**  $F$  to make things work right.
  - If not cube-free, just divide by biggest common cube to simplify  $F$ .

# Kernel Algorithm

```
FindKernels( cube-free SOP expression F ) {  
  K = empty;  
  for ( each variable x in F ) {  
    if ( there are at least 2 cubes in F that have variable x ) {  
      let S = { cubes in F that have variable x in them };  
      let co = cube that results from intersection of all cubes in S,  
              this will be the product of just those literals  
              that appear in each of these cubes in S;  
      K = K ∪ FindKernels( F / co );  
    }  
  }  
  K = K ∪ F ;  
  return( K );  
}
```

Cube-free **F** is always its own kernel, with trivial co-kernel = **1**

# Kernelling Example

```
FindKernels( F ):  
for (each var x in F ) {  
  if (x in  $\geq 2$  cubes in F) {  
    co = intersection of cubes;  
    K=K  $\cup$  FindKernels(F/co) ;  
  }  
}  
K = K  $\cup$  F ;  
return( K );
```

$$F = ace + bce + de + g$$

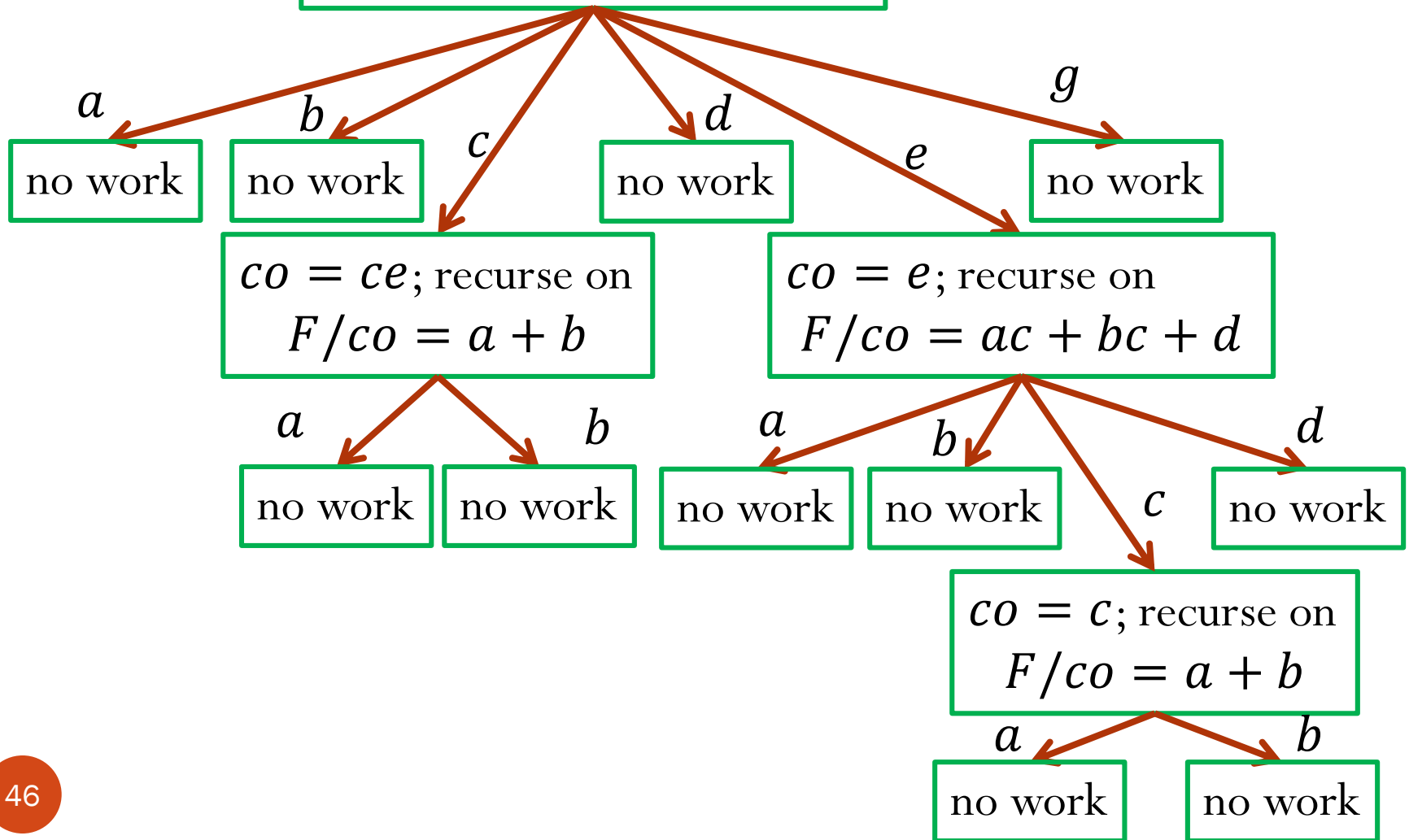
- $a$ : only 1 cube with  $a$ , no work.
- $b$ : only 1 cube with  $b$ , no work.
- $c$ : two cubes  $ace$  and  $bce$  with  $c$ .
  - $co = ace \cap bce = ce$
  - $F/co = a + b$
  - **Recurse** on  $a + b$
- $d$ : only 1 cube with  $d$ , no work.
- $e$ : three cubes  $ace$ ,  $bce$ , and  $de$  with  $e$ .
  - $co = ace \cap bce \cap de = e$
  - $F/co = ac + bc + d$
  - **Recurse** on  $ac + bc + d$
- $g$ : only 1 cube with  $g$ , no work.

# Kernelling Example (cont.)

- **Recurse** on  $a + b$ 
  - No work for variables  $a$  and  $b$ , since one cube with  $a/b$ .
- **Recurse** on  $ac + bc + d$ 
  - No work for variables  $a, b, d$ , since one cube with  $a/b/d$ .
  - $c$ : two cubes  $ac$  and  $bc$  with  $c$ .
    - $co = ac \cap bc = c$
    - $F/co = a + b$
    - **Recurse** on  $a + b$  (the same as above)

# Kernelling Example (cont.)

$$F = ace + bce + de + g$$



# Kernelling Example (cont.)

```

FindKernels( F ):
for (each var x in F) {
    ...
}
K = K ∪ F;
return( K );
    
```

$$F = ace + bce + de + g$$

$$\text{Kernels } K = \{a + b, ac + bc + d, ace + bce + de + g\}$$

$$\text{return } K = \{a + b, ac + bc + d\}$$

$$co = e; \text{ recurse on } F/co = ac + bc + d$$

$$\text{return } K = \{a + b\}$$

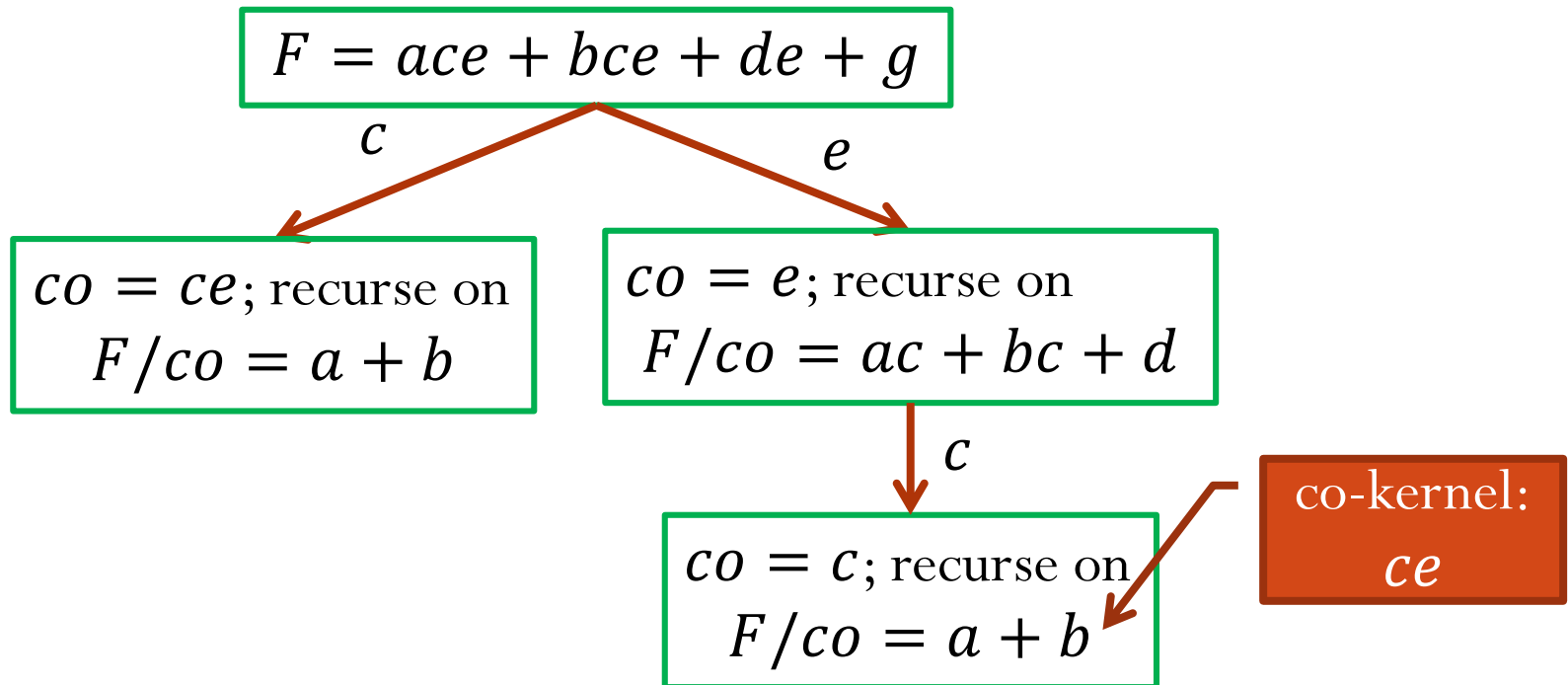
$$co = ce; \text{ recurse on } F/co = a + b$$

$$\text{return } K = \{a + b\}$$

$$co = c; \text{ recurse on } F/co = a + b$$

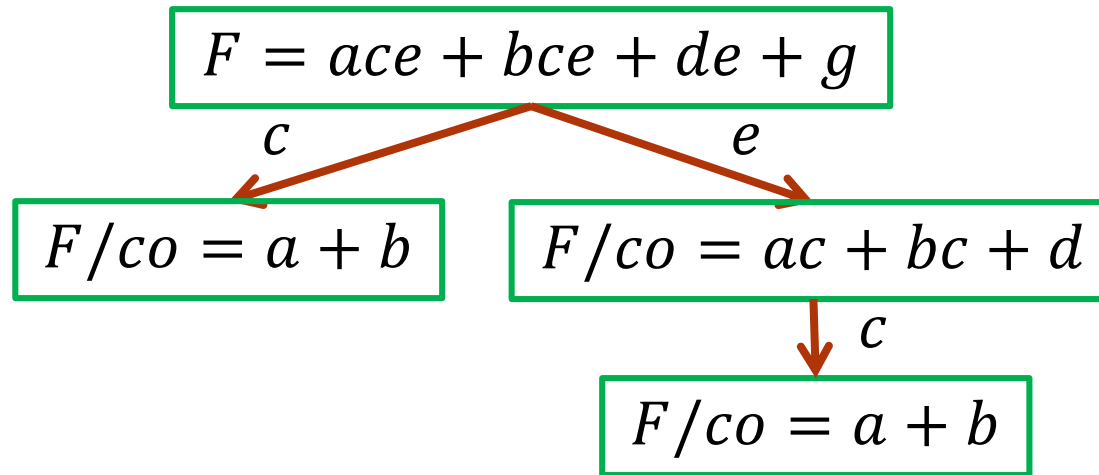
# Get Co-Kernels

- With this algorithm ...
  - Can find **all** the kernels and **co-kernels** too.
  - Get co-kernels by **ANDing** the divisor  $CO$  cubes up **recursion tree**.





# One Tiny Problem



- The algorithm will revisit same kernel **multiple** times.
  - Why? Kernel you get for co-kernel  $abc$  is same as for  $cba$ , but current algorithm **doesn't know this** and will find same kernel for both co-kernels.
- Solution: remember which variables already tried in co-kernels. A little extrabook keeping solves this.

# Multilevel Synthesis Models: Summary

- **Boolean network model**

- Like a gate network, but each node in network is an SOP form.
- Supports many operations to add, reduce, simplify nodes in network.

- **Algebraic model & algebraic division**

- Simplifies Boolean functions to behave like polynomials of reals.
- Divides one Boolean function by another:

$$F = (\text{divisor } D) \cdot (\text{quotient } Q) + \text{remainder } R$$

- **Kernels / Co-kernels** of a function F

- **Kernel** = **cube-free** quotient obtained by dividing by a single cube (**co-kernel**)
- **Intersections** of kernels of two functions give all **multiple-cube common divisors** (Brayton & McMullen theorem).

# Notes

- The **algebraic model** (and **division**) are not the only options.
  - There are also “**Boolean division**” models and algorithms that don’t lose expressivity.
  - ..but they are more complex.
  - Rich universe of models & methods here.

# Good References

- R.K. Brayton, R. Rudell, A. Sangiovanni-Vincentelli, A.R. Wang, “MIS: A Multiple-Level Logic Optimization System,” *IEEE Transactions on CAD of ICs*, vol. CAD-6, no. 6, November 1987, pp. 1062-1081.
- Giovanni De Micheli, *Synthesis and Optimization of Digital Circuits*, McGraw-Hill, 1994.

Next question:  
what are the **best** common divisors to get?

# How Do We Find Good Divisors?

- The operator is called **extraction**.
  - Want to extract either **single-cube divisor** or **multiple-cube divisor** from multiple expressions.
- How do we **extract** good divisors?
- Solution:
  - When you want a **single-cube divisor**, go look for **co-kernels**.
  - When you want a **multiple-cube divisor**, go look for **kernels**.

# Approach Overview

- For **single cube extraction**
  - Build a very large matrix of **0s and 1s**
  - Heuristically look for “**prime rectangles**” in this matrix
  - Each such “prime” is a good common single-cube divisor
- For **multiple cube extraction**
  - Build a (different) very large matrix of **0s and 1s**
  - Heuristically look for “**prime rectangles**” in this matrix
  - Each such “prime” is a good multiple-cube divisor
- Surprisingly, a lot like Karnaugh maps!
  - Except we do it all algorithmically.

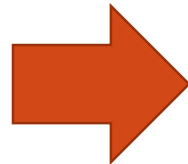
# Single Cube Extract: Matrix Representation

- **Given**: a set of SOP Boolean equations (P,Q,R).
- Construct the **cube-literal matrix** as follows:
  - One row for each unique product term.
  - One column for each unique literal.
  - A “1” in the matrix if this product term uses this literal, else a “-”.

$$P = abc + abd + eg$$

$$Q = abfg$$

$$R = bd + ef$$



		a	b	c	d	e	f	g
		1	2	3	4	5	6	7
abc	1	1	1	1	-	-	-	-
abd	2	1	1	-	1	-	-	-
eg	3	-	-	-	-	1	-	1
abfg	4	1	1	-	-	-	1	1
bd	5	-	1	-	1	-	-	-
ef	6	-	-	-	-	1	1	-

# Covering this Matrix: Prime Rectangles

- A **rectangle** of a cube-literal matrix is a set of rows R and columns C that has a '1' in every **row/column intersection**.
  - Need not be contiguous rows or columns in matrix. Any set of rows or columns is fine.

		a	b	c	d	e	f	g
	1	1	2	3	4	5	6	7
abc	1	1	1	1	-	-	-	-
abd	2	1	1	-	1	-	-	-
eg	3	-	-	-	-	1	-	1
abfg	4	1	1	-	-	-	1	1
bd	5	-	1	-	1	-	-	-
ef	6	-	-	-	-	1	1	-



# Covering this Matrix: Prime Rectangles

- A **prime rectangle** is a rectangle that cannot be made any bigger by adding another row or a column.

		a	b	c	d	e	f	g
	1	1	2	3	4	5	6	7
abc	1	1	1	1	-	-	-	-
abd	2	1	1	-	1	-	-	-
eg	3	-	-	-	-	1	-	1
abfg	4	1	1	-	-	-	1	1
bd	5	-	1	-	1	-	-	-
ef	6	-	-	-	-	1	1	-

# Prime Rectangle Columns = Divisor!

- **Primes** are “biggest possible” common single-cube divisors.
  - **Makes sense**: columns of the prime rectangle tell you the literals in the single-cube divisor, while rows tell you which product terms you can divide!

		a	b	c	d	e	f	g
		1	2	3	4	5	6	7
abc	1	1	1	1	-	-	-	-
abd	2	1	1	-	1	-	-	-
eg	3	-	-	-	-	1	-	1
abfg	4	1	1	-	-	-	1	1
bd	5	-	1	-	1	-	-	-
ef	6	-	-	-	-	1	1	-



Single-cube divisor:  
 $X = ab$

# Prime Rectangle Columns = Divisor!

		a	b	c	d	e	f	g
		1	2	3	4	5	6	7
abc	1	1	1	1	-	-	-	-
abd	2	1	1	-	1	-	-	-
eg	3	-	-	-	-	1	-	1
abfg	4	1	1	-	-	-	1	1
bd	5	-	1	-	1	-	-	-
ef	6	-	-	-	-	1	1	-



Single-cube divisor:  
 $X = ab$

$$P = abc + abd + eg$$

$$Q = abfg$$

$$R = bd + ef$$



$$P = Xc + Xd + eg$$

$$Q = Xfg$$

$$R = bd + ef$$

$$X = ab$$

# Simple Bookkeeping to Track # Literals

- Recall: we factor & extract to **reduce literals** in logic network.
  - Would be nice if there was a simple formula to compute this.
- Indeed, there is:
  - Start with a prime rectangle.
  - Let  $C = \#$  columns in rectangle.
  - For each row  $r$  in rectangle: let  $\text{Weight}(r) = \#$  times this product appears in network.
  - Compute  $L = (C - 1) \times [\sum_{\text{rows } r} \text{Weight}(r)] - C$ .
- **Nice result**: for a prime rectangle,  $L = \#$  **literals saved**
  - **To be precise**: if you count literals before extracting this single-cube divisor, and after,  $L$  is how many literals are saved.

# Compute Saved Literals: Example

$$R = abw + wz$$

$$S = abw + aby$$

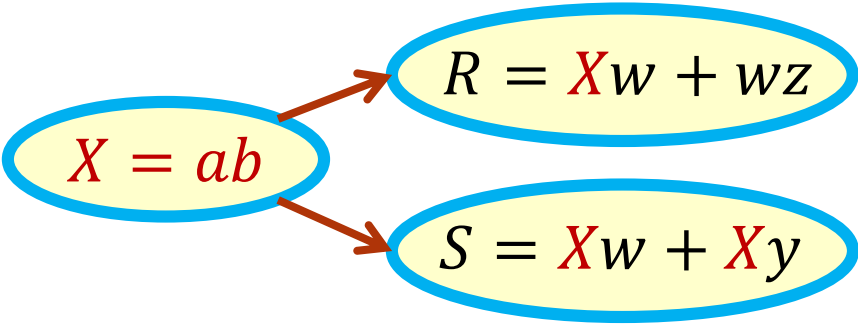


		a	b	w	y	z
	1	1	2	3	4	5
abw	1	1	1	1	-	-
wz	2	-	-	1	-	1
aby	3	1	1	-	1	-

Original # literals: 11



# saved: 1



After extraction # literals: 10

# Compute Saved Literals: Example

$$R = abw + wz$$

$$S = abw + aby$$

		a	b	w	y	z
		1	2	3	4	5
abw	1	1	1	1	-	-
wz	2	-	-	1	-	1
aby	3	1	1	-	1	-

Result by Counting:  
# saved: 1

- Now apply formula  $L = (C - 1) \times [\sum_{\text{rows } r} \text{Weight}(r)] - C$ 
  - $C = \#$  columns in rectangle  $\Rightarrow 2$
  - $\text{Weight}(abw) \Rightarrow 2$  (appear twice in the network)
  - $\text{Weight}(aby) \Rightarrow 1$  (appear once in the network)
  - $L = (2 - 1) \times (2 + 1) - 2 = 1$  **Correct!**

# How About Multiple-Cube Factors?

- Remarkably, a very similar matrix-rectangle-prime concept.
  - Make an appropriate **matrix**. Find **prime rectangle**. Do literal count **bookkeeping** with numbers associated with rows/columns.
- Given: A set of Boolean functions (nodes in a network)
$$P = af + bf + ag + cg + ade + bde + cde$$
$$Q = af + bf + ace + bce$$
$$R = ade + cde$$
- First: find **kernels** of each of these functions.
  - Why? Brayton-McMullen theorem: Multiple-cube factors are **intersections of the product terms in the kernels** for each of these functions.

# Kernels / Co-Kernels of P,Q,R Example

- $P = af + bf + ag + cg + ade + bde + cde$ 
  - Co-kernel  $a$ , kernel  $de + f + g$
  - Co-kernel  $b$ , kernel  $de + f$
  - Co-kernel  $c$ , kernel  $de + g$
  - Co-kernel  $de$ , kernel  $a + b + c$
  - Co-kernel  $f$ , kernel  $a + b$
  - Co-kernel  $g$ , kernel  $a + c$
  - Co-kernel  $1$ , kernel  $af + bf + ag + cg + ade + bde + cde$  (**trivial, ignore**)



# Kernels / Co-Kernels of P,Q,R Example

- $Q = af + bf + ace + bce$ 
  - Co-kernel  $a$ , kernel  $ce + f$
  - Co-kernel  $b$ , kernel  $ce + f$
  - Co-kernel  $ce$ , kernel  $a + b$
  - Co-kernel  $f$ , kernel  $a + b$
  - Co-kernel  $1$ , kernel  $af + bf + ace + bce$  **(trivial, ignore)**
  
- $R = ade + cde$ 
  - Co-kernel  $de$ , kernel  $a + c$
  - Note:  $R$  is not its own kernel, why?

# New Matrix: Co-Kernel-Cube Matrix

- One row for each **unique** (function, co-kernel) **pair** in problem.
- One column for each **unique cube** among all kernels in problem.

$P$ : co-kernel  $a$ , kernel  $de + f + g$

$P$ : co-kernel  $b$ , kernel  $de + f$

$P$ : co-kernel  $c$ , kernel  $de + g$

$P$ : co-kernel  $de$ , kernel  $a + b + c$

$P$ : co-kernel  $f$ , kernel  $a + b$

$P$ : co-kernel  $g$ , kernel  $a + c$

$Q$ : co-kernel  $a$ , kernel  $ce + f$

$Q$ : co-kernel  $b$ , kernel  $ce + f$

$Q$ : co-kernel  $ce$ , kernel  $a + b$

$Q$ : co-kernel  $f$ , kernel  $a + b$

$R$ : co-kernel  $de$ , kernel  $a + c$

			a	b	c	ce	de	f	g
			1	2	3	4	5	6	7
P	a	1							
P	b	2							
P	c	3							
P	de	4							
P	f	5							
P	g	6							
Q	a	7							
Q	b	8							
Q	ce	9							
Q	f	10							
R	de	11							

?

# Entries in the Co-Kernel-Cube Matrix

- For each **row**, take the co-kernel, go look at the associated kernel.
- Look at **cubes** in this kernel: put “1” in columns that are cubes in this kernel; else put “-”

*P*: co-kernel *a*, kernel  $de + f + g$

*P*: co-kernel *b*, kernel  $de + f$

*P*: co-kernel *c*, kernel  $de + g$

*P*: co-kernel  $de$ , kernel  $a + b + c$

*P*: co-kernel *f*, kernel  $a + b$

*P*: co-kernel *g*, kernel  $a + c$

*Q*: co-kernel *a*, kernel  $ce + f$

*Q*: co-kernel *b*, kernel  $ce + f$

*Q*: co-kernel  $ce$ , kernel  $a + b$

*Q*: co-kernel *f*, kernel  $a + b$

*R*: co-kernel  $de$ , kernel  $a + c$

			a	b	c	ce	de	f	g
			1	2	3	4	5	6	7
P	a	1	-	-	-	-	1	1	1
P	b	2	-	-	-	-	1	1	-
P	c	3	-	-	-	-	1	-	1
P	de	4	1	1	1	-	-	-	-
P	f	5	1	1	-	-	-	-	-
P	g	6	1	-	1	-	-	-	-
Q	a	7	-	-	-	1	-	1	-
Q	b	8	-	-	-	1	-	1	-
Q	ce	9	1	1	-	-	-	-	-
Q	f	10	1	1	-	-	-	-	-
R	de	11	1	-	1	-	-	-	-

# Entries in the Co-Kernel-Cube Matrix

- Each row gives the kernel of the function (e.g.,  $P$ ) obtained by dividing the co-kernel (e.g.,  $a$ ).

$P$ : co-kernel  $a$ , kernel  $de + f + g$

			a	b	c	ce	de	f	g
			1	2	3	4	5	6	7
P	a	1	-	-	-	-	1	1	1
P	b	2	-	-	-	-	1	1	-
P	c	3	-	-	-	-	1	-	1
P	de	4	1	1	1	-	-	-	-
P	f	5	1	1	-	-	-	-	-
P	g	6	1	-	1	-	-	-	-
Q	a	7	-	-	-	1	-	1	-
Q	b	8	-	-	-	1	-	1	-
Q	ce	9	1	1	-	-	-	-	-
Q	f	10	1	1	-	-	-	-	-
R	de	11	1	-	1	-	-	-	-

# Prime Rectangles in Co-Kernel-Cube Matrix

- Prime rectangle is again a good divisor: now multiple cube
  - Create the multiple cube divisor as **sum** (OR) of cubes of prime rectangle **columns**.

		a	b	c	ce	de	f	g
		1	2	3	4	5	6	7
P	a	1	-	-	-	1	1	1
P	b	2	-	-	-	1	1	-
P	c	3	-	-	-	1	-	1
P	de	4	1	1	1	-	-	-
P	f	5	1	1	-	-	-	-
P	g	6	1	-	1	-	-	-
Q	a	7	-	-	-	1	-	1
Q	b	8	-	-	-	1	-	1
Q	ce	9	1	1	-	-	-	-
Q	f	10	1	1	-	-	-	-
R	de	11	1	-	1	-	-	-



**(a+b)** is the multiple cube divisor!

$P = (de) \cdot (a+b + \text{stuff1}) + \text{rem1}$

$P = (f) \cdot (a+b + \text{stuff2}) + \text{rem2}$

$Q = (ce) \cdot (a+b + \text{stuff3}) + \text{rem3}$

$Q = (f) \cdot (a+b + \text{stuff4}) + \text{rem4}$

# Simple Formula to Get # Literals Saved

- For each column  $c$  in rectangle: let  $\text{Weight}(c) = \#$  literals in column cube.
- For each row  $r$  in rectangle: let  $\text{Weight}(r) = 1 + \#$  literals in co-kernel label.
- For each “1” covered at row  $r$  and column  $c$ : AND row co-kernel and column cube; let  $\text{Value}(r, c) = \#$  literals in this new ANDed product.
- **# literals saved** is

$$\begin{aligned} & L \\ &= \sum_{\text{row } r} \sum_{\text{col } c} \text{Value}(r, c) - \sum_{\text{row } r} \text{Weight}(r) \\ & \quad - \sum_{\text{col } c} \text{Weight}(c) \end{aligned}$$

# Compute Saved Literals: Example

$$P = af + bf + ag + cg + ade + bde + cde$$

$$Q = af + bf + ace + bce$$

$$R = ade + cde$$

Original # literals: 33

Build Matrix

Extraction

		a	b	c	ce	de	f	g
		1	2	3	4	5	6	7
P	a	1	-	-	-	-	1	1
P	b	2	-	-	-	-	1	1
P	c	3	-	-	-	-	1	-
P	de	4	1	1	1	-	-	-
P	f	5	1	1	-	-	-	-
P	g	6	1	-	1	-	-	-
Q	a	7	-	-	-	1	-	1
Q	b	8	-	-	-	1	-	1
Q	ce	9	1	1	-	-	-	-
Q	f	10	1	1	-	-	-	-
R	de	11	1	-	1	-	-	-

$$P = Xf + Xde + ag + cg + cde$$

$$X = a + b$$

$$Q = Xf + Xce$$

$$R = ade + cde$$

After extraction # literals: 25

# saved: 8

# Compute Saved Literals: Example

		a	b	c	ce	de	f	g
		1	2	3	4	5	6	7
P	a	1	-	-	-	-	1	1
P	b	2	-	-	-	-	1	1
P	c	3	-	-	-	-	1	-
P	de	4	1	1	1	-	-	-
P	f	5	1	1	-	-	-	-
P	g	6	1	-	1	-	-	-
Q	a	7	-	-	-	1	-	1
Q	b	8	-	-	-	1	-	1
Q	ce	9	1	1	-	-	-	-
Q	f	10	1	1	-	-	-	-
R	de	11	1	-	1	-	-	-

$$P = af + bf + ag + cg + ade + bde + cde$$

$$Q = af + bf + ace + bce$$

$$R = ade + cde$$

# saved: 8

- Column weight
  - $\text{Weight}(a) = \#\text{literals in "a"} \Rightarrow 1$
  - $\text{Weight}(b) = \#\text{literals in "b"} \Rightarrow 1$
- Row weight
  - $\text{Weight}((P, de)) = 1 + \#\text{literals in "de"} \Rightarrow 3$
  - $\text{Weight}((P, f)) = 1 + \#\text{literals in "f"} \Rightarrow 2$
  - $\text{Weight}((Q, ce)) = 1 + \#\text{literals in "ce"} \Rightarrow 3$
  - $\text{Weight}((Q, f)) = 1 + \#\text{literals in "f"} \Rightarrow 2$



# Compute Saved Literals: Example

		a	b	c	ce	de	f	g
		1	2	3	4	5	6	7
P	a	1	-	-	-	1	1	1
P	b	2	-	-	Values		-	-
P	c	3	-	-			-	1
P	de	4	1	1	3	3	-	-
P	f	5	1	1	2	2	-	-
P	g	6	1	-	-	-	-	-
Q	a	7	-	-	1	-	1	-
Q	b	8	-	-	1	-	1	-
Q	ce	9	1	1	3	3	-	-
Q	f	10	1	1	2	2	-	-
R	de	11	1	-	1	-	-	-

# saved: 8

- Column weight
  - Weight(a) = 1; Weight(b) = 1
- Row weight
  - Weight((P, de)) = 3; Weight((P, f)) = 2
  - Weight((Q, ce)) = 3; Weight((Q, f)) = 2
- Value(r,c): # literals in the **product** of **row co-kernel** and **column cube**.
- Apply formula  $L =$ 

$$\sum_{\text{row } r} \sum_{\text{col } c} \text{Value}(r, c) - \sum_{\text{row } r} \text{Weight}(r) - \sum_{\text{col } c} \text{Weight}(c)$$

$$= 20 - 10 - 2 = 8$$

Correct!

# Details for Both Single/Multiple Cube Extraction

- You can extract a **second**, **third**, etc., divisor using same matrix.
  - Works for both single-cube and multiple-cube divisors.
- ...but must **update** matrix to reflect new Boolean logic network.
  - Because the node contents are different, and there is a new divisor node in network.
  - For multiple-cube case, must **kernel** new divisor nodes to update matrix.
  - All mechanical. A bit tedious. Just skip it...
  - For us: just know how to **extract first good divisor** is good enough.

# How to Find Prime Rectangle in Matrix?

- **Greedy heuristics** work well for this rectangle covering problem.
  - Start with a single row rectangle with “good #literal savings”.
  - Grow the rectangle alternatively by adding more rows, more columns.
- Example: **Rudell’s Ping Pong heuristic**.
  - From his Berkeley PhD dissertation in 1989.
  - **Very good** heuristic:
    - **< 1%** of optimal result.
    - **10~100x faster** than brute force approach.

# Rudell's Ping Pong Heuristic

1. Pick the **best single row** (the 1-row rectangle with best #literals saved).
2. Look at other rows **with 1s in same places** (may have more 1s). Add the one that **maximizes** #literals saved. Iterate until can't find any more.
3. Look at other columns **with 1s in same places** (may have more 1s). Add the one that **maximizes** # literals saved. Iterate until can't find any more.
4. Go to 2.
5. Quit when can't grow rectangle any more in any direction.

# Extraction: Summary

- **Single cube extraction**
  - Build the cube-literal matrix.
  - Each prime rectangle is a good **single cube divisor**.
  - Simple bookkeeping lets us obtain savings in #literals.
- **Multiple cube extraction**
  - Kernel all the expressions in network; build the co-kernel-cube matrix.
  - Each prime rectangle is a good **multiple cube divisor**.
  - Simple bookkeeping lets us obtain savings in #literals.
- Mechanically, both are **rectangle covering** problems (very like Karnaugh maps!)
  - Good **heuristics** to obtain a good prime rectangle, fast and effective.

# Aside: How to We Really Do This?

- Do **not** use rectangle covering on **all** kernels/co-kernels
  - Too expensive to do rectangle problem on big circuits (>20K gates)
  - Too expensive to go compute **complete** set of kernels, co-kernels
- Often use heuristics to find a “**quick**” set of likely divisors.
  - Don't fully kernel each node of network: too many cubes to consider. Instead, can extract a **subset** of useful kernels quickly.
  - Then, can either do rectangle cover on these smaller problems (smaller since fewer things to consider in covering problem)...
  - ...or, try to do simpler overall network restructuring, e.g., try all pairwise **substitutions** of one node into another node: keep good ones, continue in a greedy way.

# References

- R.K. Brayton, R. Rudell, A. Sangiovanni-Vincentelli, A.R. Wang, "MIS: A Multiple-Level Logic Optimization System," *IEEE Transactions on CAD of ICs*, vol. CAD-6, no. 6, November 1987, pp. 1062-1081.
- Giovanni De Micheli, *Synthesis and Optimization of Digital Circuits*, McGraw-Hill, 1994.
- R.K. Brayton, R. Rudell, A.S. Vincentelli, and A. Wang, "Multi-Level Logic Optimization and the Rectangular Covering Problem," *Proceedings of the International Conference on Computer Aided Design*, pp. 66-69, 1987.
- Richard Rudell, *Logic Synthesis for VLSI Design*, PhD Thesis, Dept of EECS, University of California at Berkeley, 1989.
- Srinivas Devadas, Abhijit Gosh, Kurt Keutzer, *Logic Synthesis*, McGraw Hill Inc., 1994.

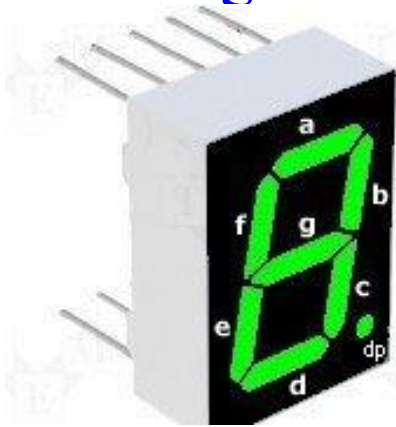
# Don't Cares

- We made progress on multi-level logic by **simplifying** the model.
  - Algebraic model: we **get rid of** a lot of “difficult” Boolean behaviors.
  - But we lost some optimality in the process.
- How do we put it back? One surprising answer: **Don't cares**
  - To help this, **extract** don't cares from “surrounding logic,” use them **inside each node**.
- The big difference in multi-level logic
  - Don't cares happen as a natural byproduct of Boolean network model: called **Implicit Don't Cares**.
  - They are all over the place, in fact. Very useful for simplification.
  - But they are **not explicit**. We have to **go hunt for them**...



# Don't Cares Review: 2-Level

- In basic digital design...
  - Don't Care (DC) = an input pattern that **can never happen** or you don't care the output if it happens.
  - Example: use **binary-coded decimals (BCD)** to control **seven-segment digital tube**.



How about input  $(x,y,z,w)$   
 $= (1,0,1,0), (1,0,1,1) \dots?$

Don't care!

x y z w	decimal value	segment a
0 0 0 0	0	1
0 0 0 1	1	0
0 0 1 0	2	1
0 0 1 1	3	1
0 1 0 0	4	0
0 1 0 1	5	1
0 1 1 0	6	1
0 1 1 1	7	1
1 0 0 0	8	1
1 0 0 1	9	1

# Don't Cares Review: 2-Level

- Since patterns  $(x,y,z,w)=(1,0,1,0)$ ,  $(1,0,1,1)$ ,  $(1,1,0,0)$ ,  $(1,1,0,1)$ ,  $(1,1,1,0)$ ,  $(1,1,1,1)$  are don't cares, we are **free to decide** whether  $F=1$  or  $0$ , to better **optimize**  $F$ .

x y z w	decimal value	segment a
0 0 0 0	0	1
0 0 0 1	1	0
0 0 1 0	2	1
0 0 1 1	3	1
0 1 0 0	4	0
0 1 0 1	5	1
0 1 1 0	6	1
0 1 1 1	7	1
1 0 0 0	8	1
1 0 0 1	9	1

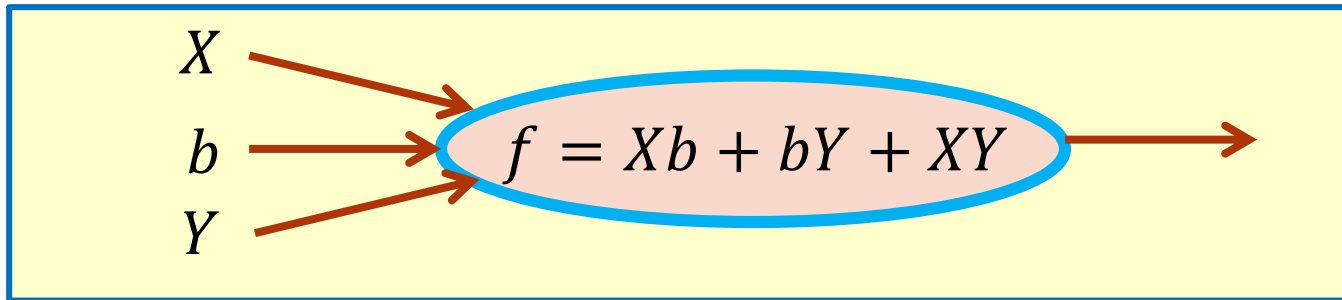
zw \ xy	00	01	11	10
00	1	0	d	1
01	0	1	d	1
11	1	1	d	d
10	1	1	d	d

# Don't Cares (DCs): Multi-level

- What's different in multi-level?
  - DCs arise **implicitly**, as a result of the **Boolean logic network structure**.
  - We must go find these implicit don't cares – we must search for them explicitly.

# Multi-level DCs: Informal Tour

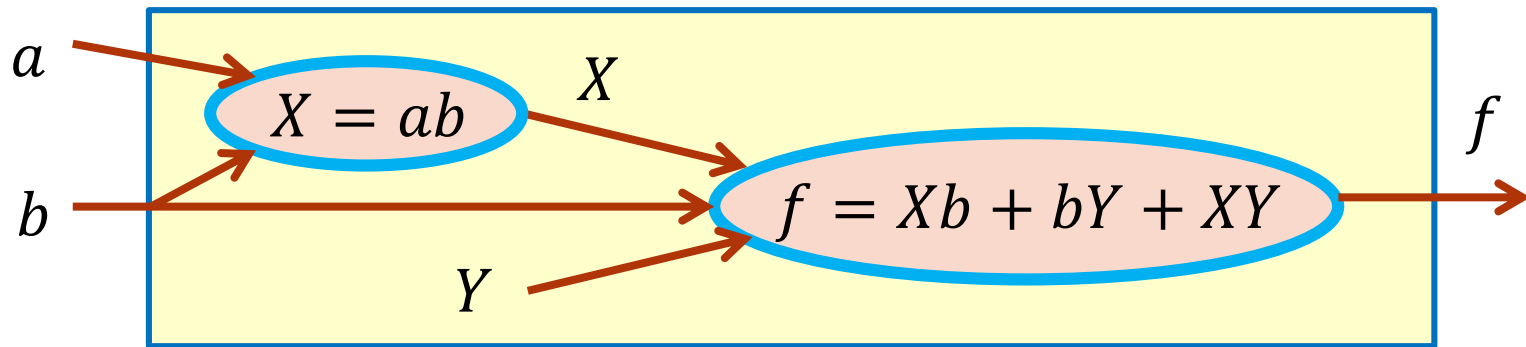
- Suppose we have a Boolean network and a node  $f$  in the network.



- Can we say anything about **don't cares** for node  $f$ ?
  - No. We don't know any "context" for surrounding parts of network.
  - As far as we can tell, all patterns of inputs  $(X, b, Y)$  are possible.
  - We **cannot further simplify** the expression for  $f$ .

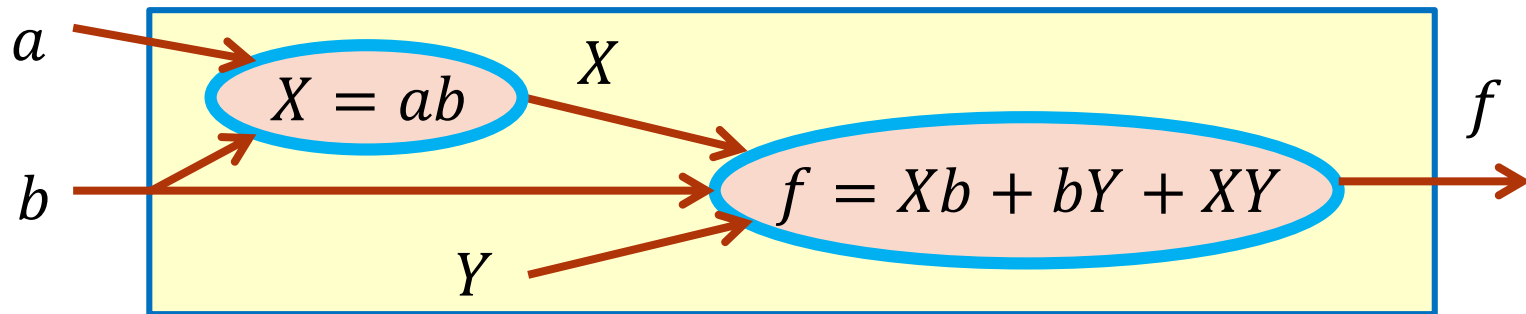
# Multi-level DCs: Informal Tour

- Now suppose we know something about input  $X$  to  $f$ :
  - Node  $X = ab$ .
  - Also assume  $a$  and  $b$  are **primary inputs (PIs)** and  $f$  is **primary output (PO)**.



- Now can we say something about DCs for node  $f$ ...?
  - **YES!**
  - Because there are some **impossible patterns** of  $(X, b, Y)$ .

# Multi-level DCs: Informal Tour



The possible input/output patterns for node  $X$

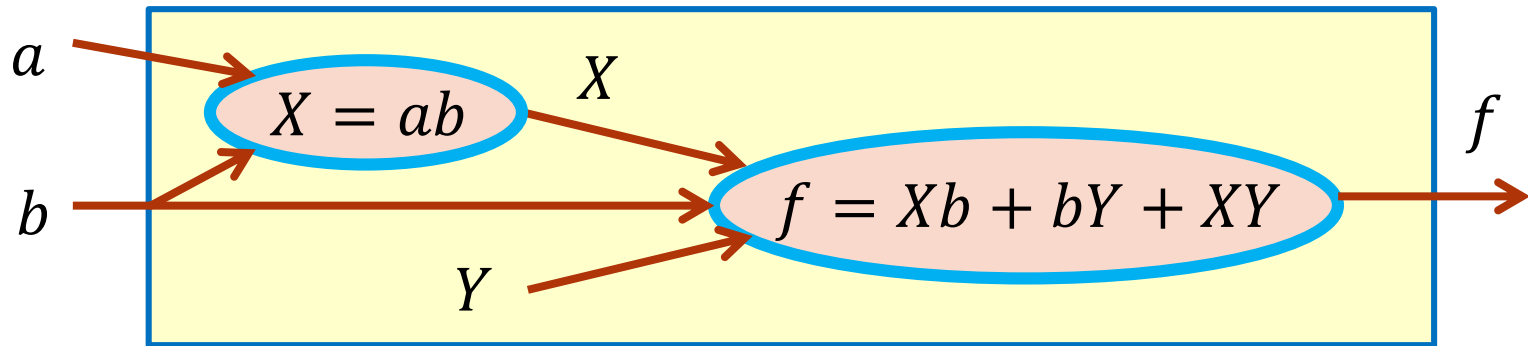
a	b	X	Can it occur?
0	0	0	Yes
0	0	1	No
0	1	0	Yes
0	1	1	No
1	0	0	Yes
1	0	1	No
1	1	0	No
1	1	1	Yes



b	X	Can it occur?
0	0	Yes
0	1	No
1	0	Yes
1	1	Yes

Impossible patterns for  $(X, b, Y)$  are:  
 $(1, 0, 0)$  and  $(1, 0, 1)$

# Multi-level DCs: Informal Tour



- Impossible patterns for  $(X, b, Y)$  are  $(1, 0, 0)$  and  $(1, 0, 1)$ .
  - With them, we can simplify  $f$ .

Can be simplified as  
 $f = X + bY$

Kmap for  $f = Xb + bY + XY$

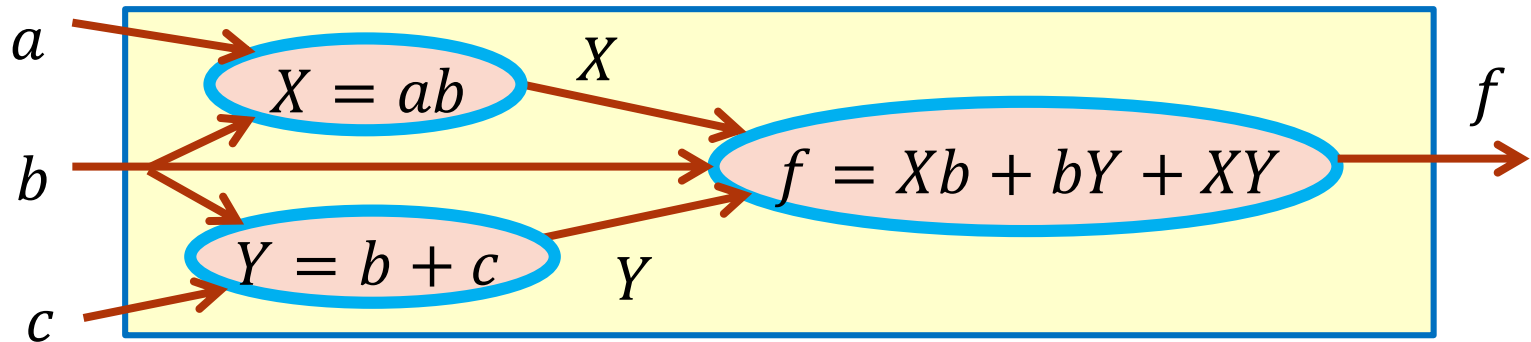
	Xb			
Y	00	01	11	10
0			1	
1		1	1	1



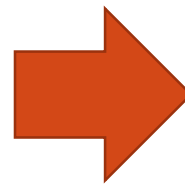
	Xb			
Y	00	01	11	10
0			1	d
1		1	1	d

# Multi-level DCs: Informal Tour

- Now further suppose  $Y = b + c$ . What will happen?



b	c	Y	Can it occur?
0	0	0	Yes
0	0	1	No
0	1	0	No
0	1	1	Yes
1	0	0	No
1	0	1	Yes
1	1	0	No
1	1	1	Yes

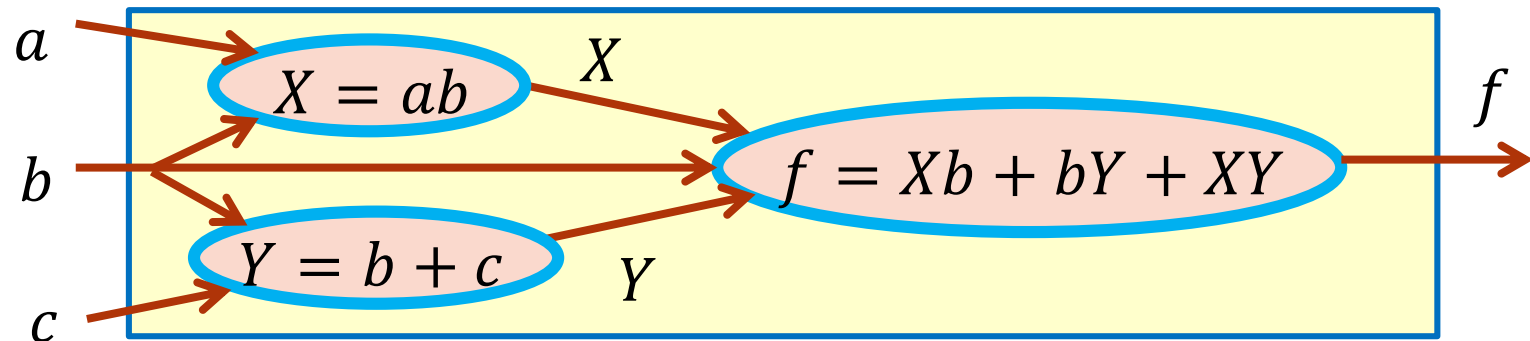


b	Y	Can it occur?
0	0	Yes
0	1	Yes
1	0	No
1	1	Yes

Impossible patterns for  $(X, b, Y)$  are:  
 $(0, 1, 0)$  and  $(1, 1, 0)$



# Multi-level DCs: Informal Tour



- Impossible patterns for  $(X, b, Y)$  are
  - $(1, 0, 0), (1, 0, 1)$  (From  $X = ab$ )
  - $(0, 1, 0), (1, 1, 0)$  (From  $Y = b + c$ )

Kmap for  $f = Xb + bY + XY$

		Xb			
Y		00	01	11	10
0				1	
1			1	1	1

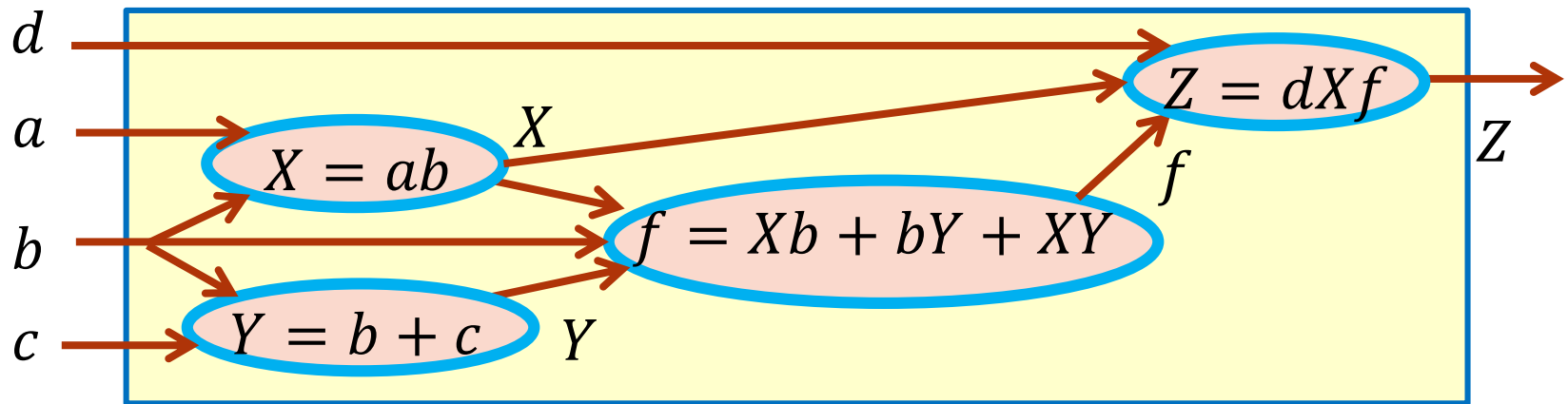


		Xb			
Y		00	01	11	10
0			d	d	d
1			1	1	d

$f$  can be simplified  
as  $f = b$

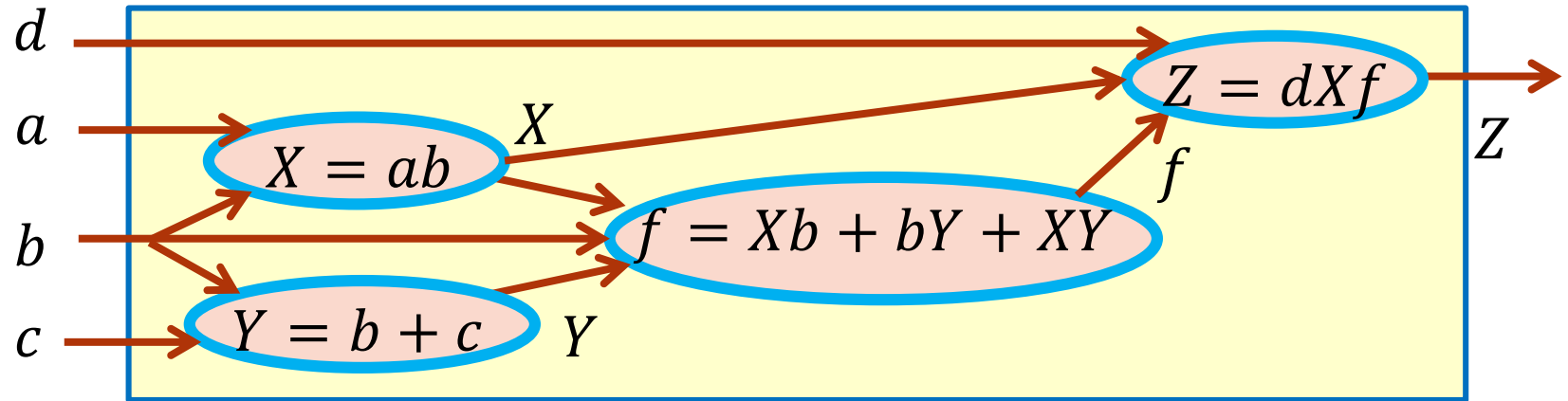
# Multi-level DCs: Informal Tour

- Now suppose  $f$  is not a **primary output**,  $Z$  is.



- **Question**: when does the value of the output of node  $f$  actually **affect** the primary output  $Z$ ?
  - Or, said **conversely**: When does it **not matter** what  $f$  is?
  - Let's go look at patterns of  $(f, X, d)$  at node  $Z$ ...

# When Is Z “Sensitive” to Value of f?

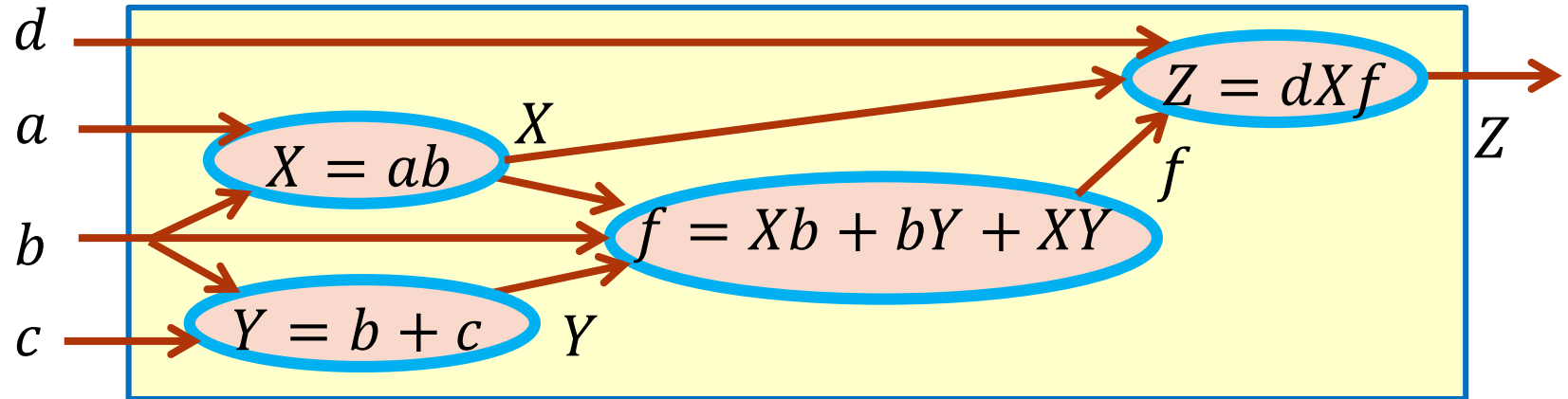


f	X	d	Z	Does f affect Z?
0	0	0	0	No
1	0	0	0	No
0	0	1	0	No
1	0	1	0	No
0	1	0	0	No
1	1	0	0	No
0	1	1	0	No
1	1	1	1	Yes

Can we use this information to find new patterns of  $(X, b, Y)$  to help us simplify  $f$  further?

**YES!**

# When Is Z “Sensitive” to Value of f?



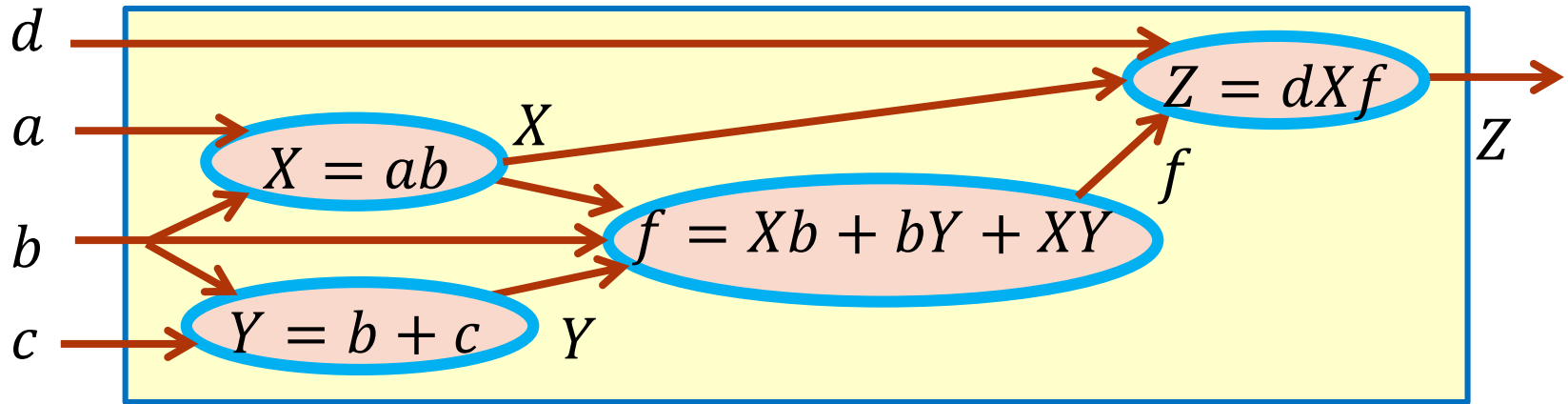
f	X	d	Z	Dose f affect Z?
0	0	0	0	No
1	0	0	0	No
0	0	1	0	No
1	0	1	0	No
0	1	0	0	No
1	1	0	0	No
0	1	1	0	Yes
1	1	1	1	Yes

What patterns at **input to  $f$**  node (i.e.,  $(X, b, Y)$ ) are DCs, because those patterns make  $Z$  output **insensitive** to changes in  $f$ ?

$$(X, b, Y) = (0, -, -)$$

This means when  $X = 0$ , we can set  $f$  to any value – it **won't change**  $Z$ . So  $(X, b, Y) = (0, -, -)$  is DC of  $f$ !

# Multi-level DCs: Informal Tour



- So, we can use this **new** DC pattern  $(0, -, -)$  to simplify  $f$  further...
  - ... with previous DC patterns  $(1,0,0)$ ,  $(1,0,1)$ ,  $(0,1,0)$ ,  $(1,1,0)$ .

Kmap for  $f = Xb + bY + XY$

		Xb			
Y		00	01	11	10
0				1	
1			1	1	1

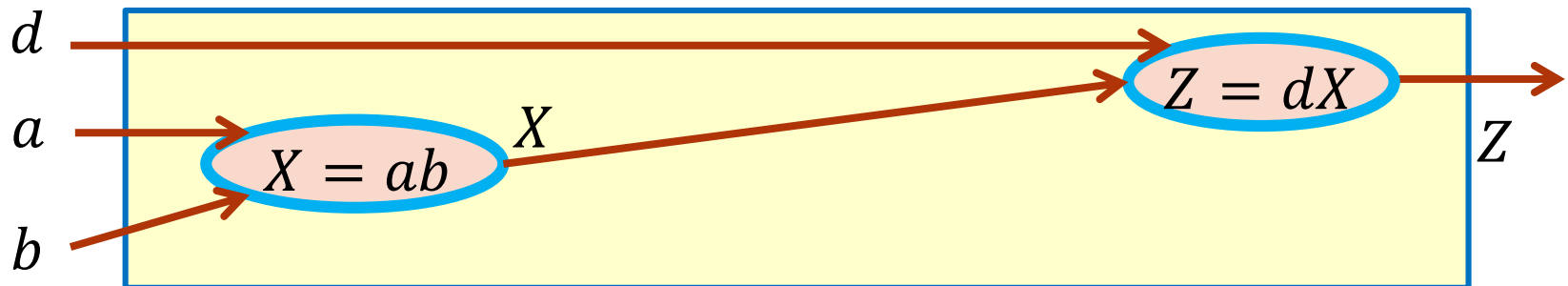
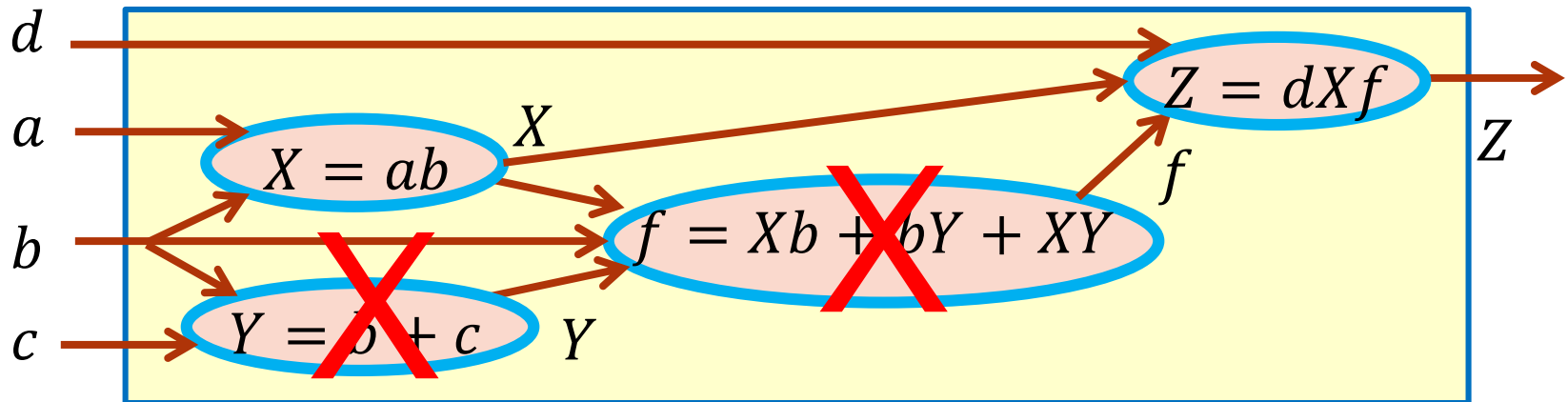


		Xb			
Y		00	01	11	10
0		d	d	d	d
1		d	1	1	d

$f$  simplified as 1

# Final Result: Multi-level DC Tour

- What happened to  $f$ ?
  - Due to network **context**, it disappeared ( $f = 1$ )!



# Summary

- Don't Cares are **implicit** in the Boolean network model.
  - They arise from the **graph structure** of the multilevel Boolean network model itself.
- Implicit Don't Cares are **powerful**.
  - They can greatly help simplify the 2-level SOP structure of any node.
- Implicit Don't Cares require **computational work** to find.
  - For this example, we just “stared at the logic” to find the DC patterns.
  - We need some **algorithms** to do this automatically!
  - This is what we need to study next ...

# Multi-Level Don't Cares

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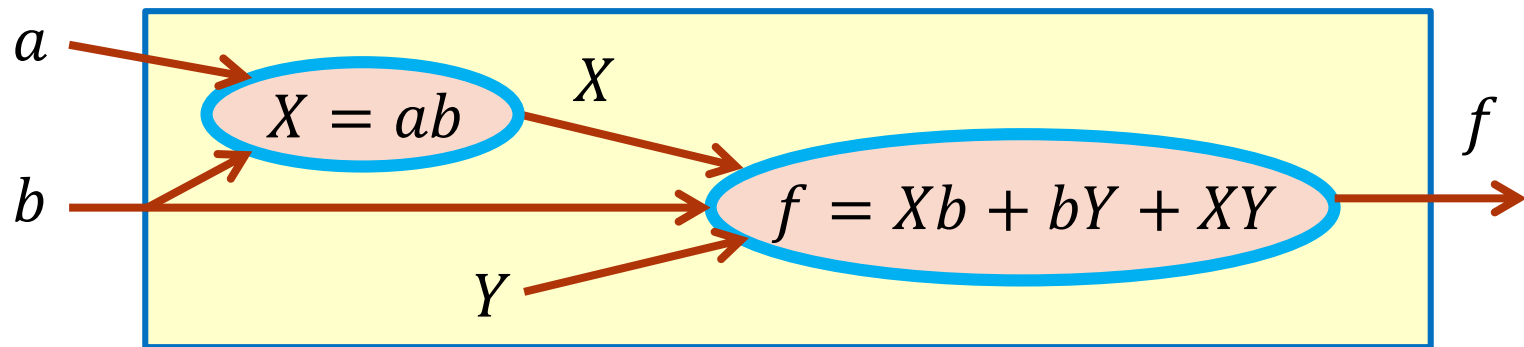


# 3 Types of Implicit DCs

- **Satisfiability** don't cares: **SDCs**
  - Belong to the **wires** inside the Boolean logic network.
  - Used to compute **controllability** don't cares (below).
- **Controllability** don't cares: **CDCs**
  - Patterns that **cannot happen at inputs** to a network node.
- **Observability** don't cares: **ODCs**
  - Patterns that “**mask**” outputs.

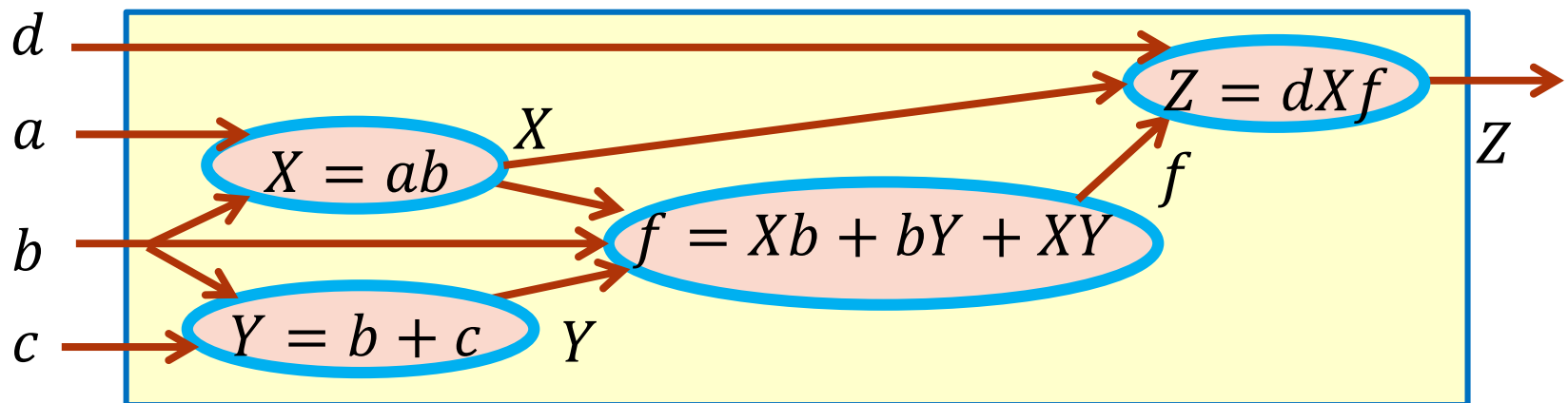
# Controllability don't cares: CDCs

- Patterns that **cannot happen at inputs** to a network node.
- **Example**
  - For node  $f$ ,  $(X, b, Y) = (1,0,0), (1,0,1)$  are CDCs.



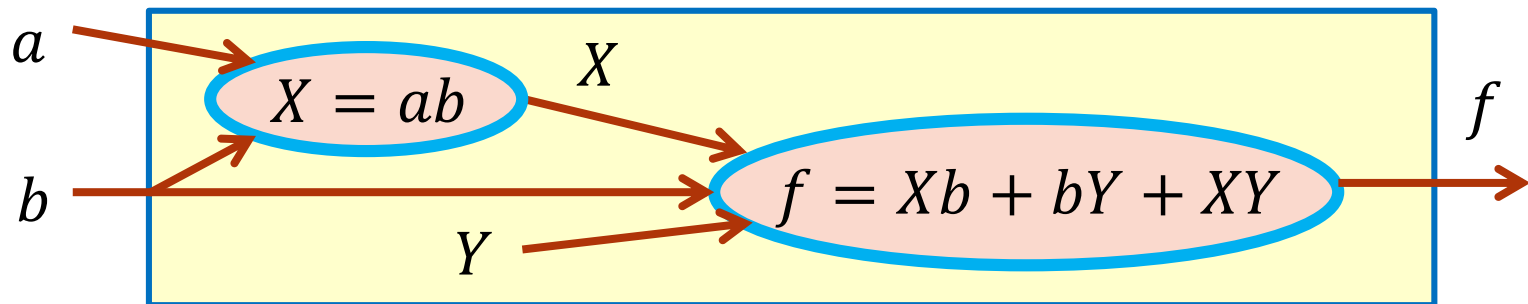
# Observability don't cares: ODCs

- Input patterns to node that make primary outputs **insensitive** to output of the node.
  - Patterns that “**mask**” outputs.
- **Example**
  - For node  $f$ ,  $(X, b, Y) = (0, -, -)$  is ODC.



# Background: Representing DC Patterns

- How shall we **represent** DC patterns at a node?
  - **Answer:** As a **Boolean function** that makes a 1 when the inputs are **these DCs**.
  - This is often called a **Don't Care Cover**.



Don't care pattern of  $(X,b,Y)=(1,0,0), (1,0,1)$

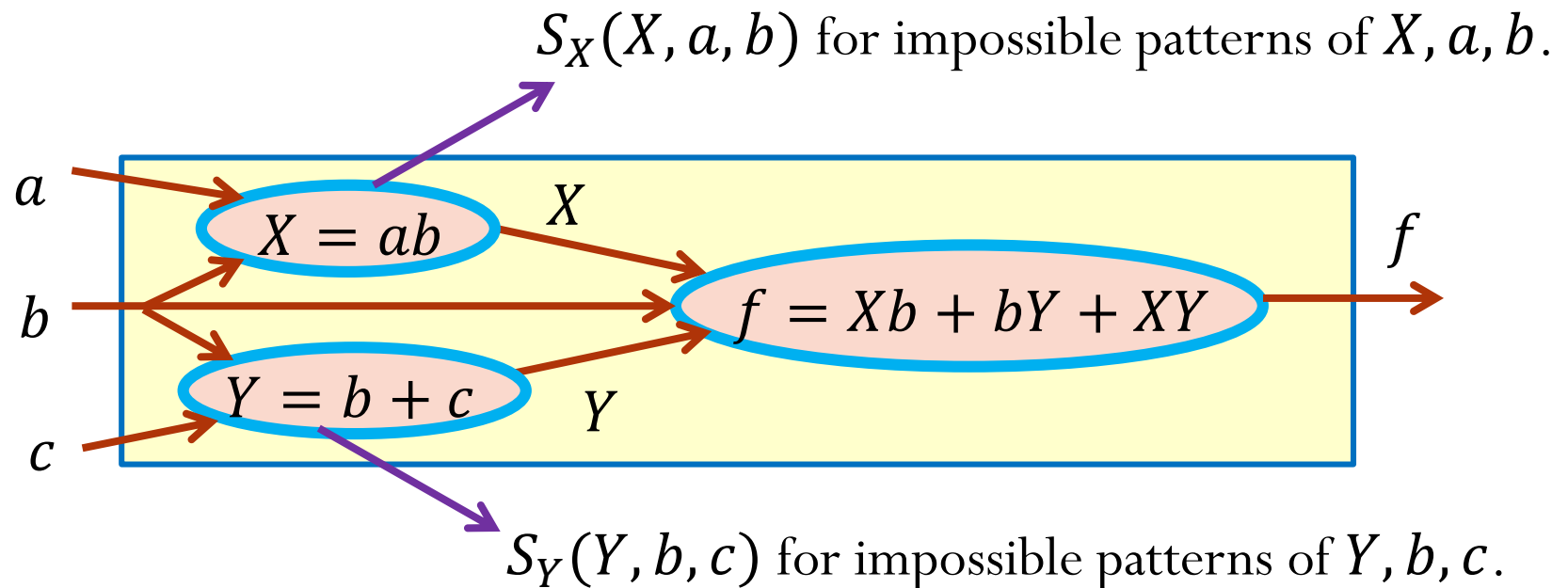
The don't care cover is  $X\bar{b}\bar{Y} + X\bar{b}Y = X\bar{b}$

# Background: Representing DC Patterns

- So, each SDC, CDC, ODC is really just another Boolean function, in this strategy.
- Why do it like this?
  - Because we can use all the other **computational Boolean algebra** techniques we know (e.g., BDDs), to **solve** for, and to **manipulate** the DC patterns.
  - This turns out to be hugely important to making the computation practical.

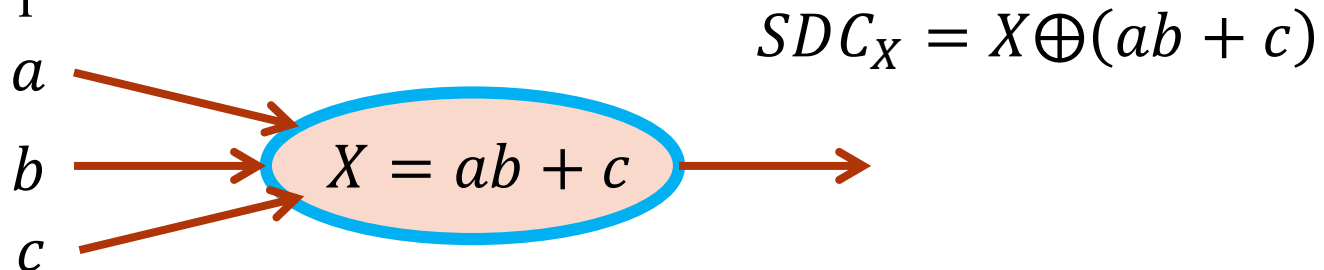
# SDCs: They “Belong” to the Wires

- One SDC for every **internal wire** in Boolean logic network.
  - The SDC represents **impossible** patterns of **inputs to, and output of**, each node.
  - If the node function is  $F$ , with inputs  $a, b, c$ , write as:  
 $S_F(F, a, b, c)$ .

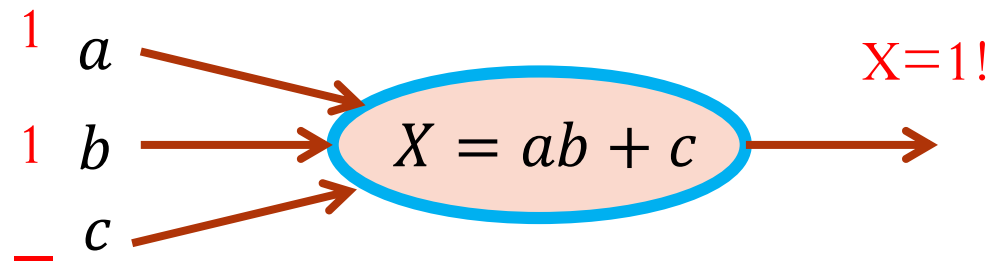


# SDCs: How to Compute

- Compute an SDC for each output wire from each internal Boolean node.
- You want an expression that is 1 when output  $X$  **does not equal** the Boolean expression for  $X$ .
  - This is just:  $X \oplus$  (expression for  $X$ )
    - Note #1: expression for  $X$  doesn't have  $X$  in it!
    - Note #2: this is the **complement** of the gate consistency function from SAT.
- Example



# SDCs: Example



- $SDC_X = X \oplus (ab + c) = \bar{X}ab + \bar{X}c + X\bar{a}\bar{c} + X\bar{b}\bar{c}$



One **impossible pattern**:  $Xabc = 011 -$

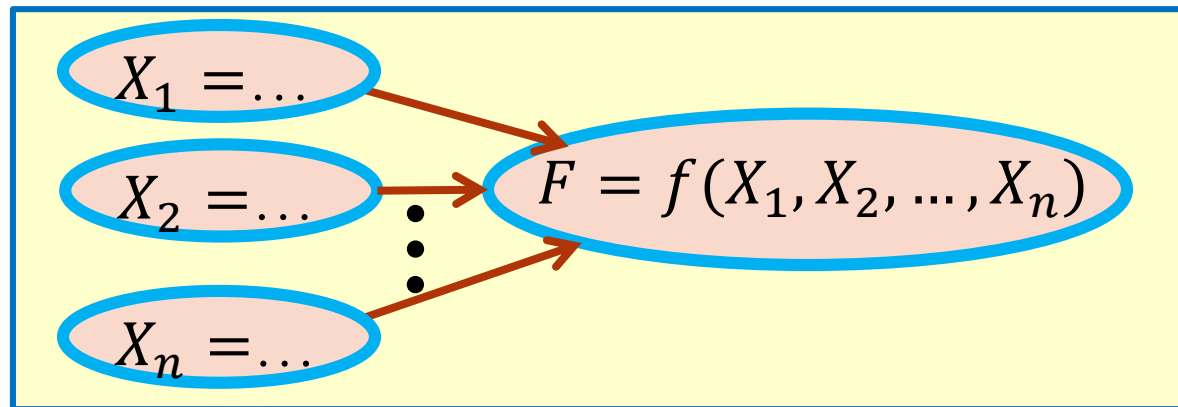


# SDCs: Summary

- SDCs are associated with every **internal wire** in Boolean logic network.
  - SDCs explain **impossible patterns** of input to, and output of, each node.
  - SDCs are easy to compute.
- SDCs alone are **not** the Don't Cares used to simplify nodes.
  - We use SDCs to **build CDCs**, which give impossible patterns at input of nodes.

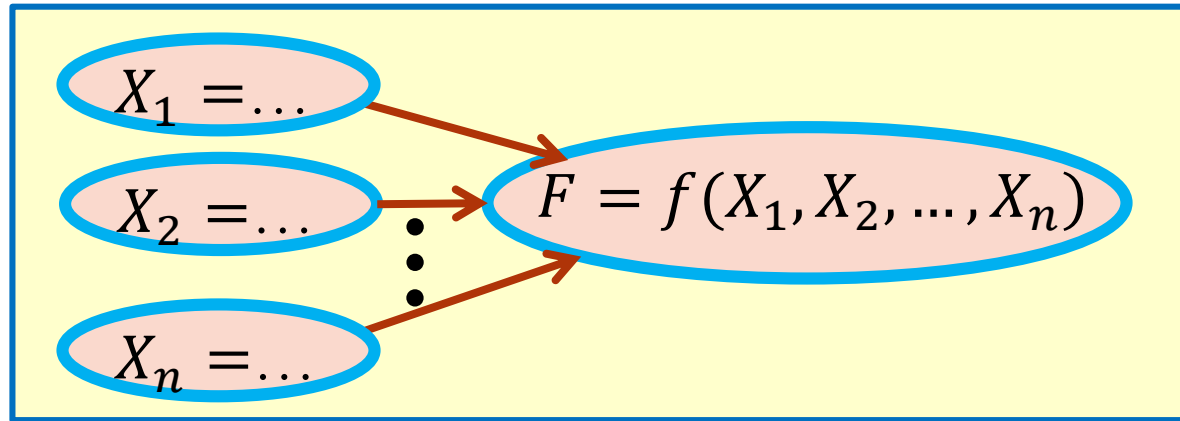
# How to Compute CDCs?

- Computational recipe:
  1. Get all the **SDCs** on the wires **input to** this node in Boolean logic network.
  2. **OR** together all these SDCs.
  3. **Universally Quantify** away all variables that are **NOT** used inside this node.



$$CDC_F(X_1, \dots, X_n) = (\forall \text{ vars not used in } F) \left[ \sum_{\text{input } X_i \text{ to } F} SDC_{X_i} \right]$$

# How to Compute CDCs?



$$CDC_F(X_1, \dots, X_n) = (\forall \text{ vars not used in } F) \left[ \sum_{\text{input } X_i \text{ to } F} SDC_{X_i} \right]$$

- **Result:** Inputs that let  $CDC_F = 1$  are **impossible patterns** at input to node!

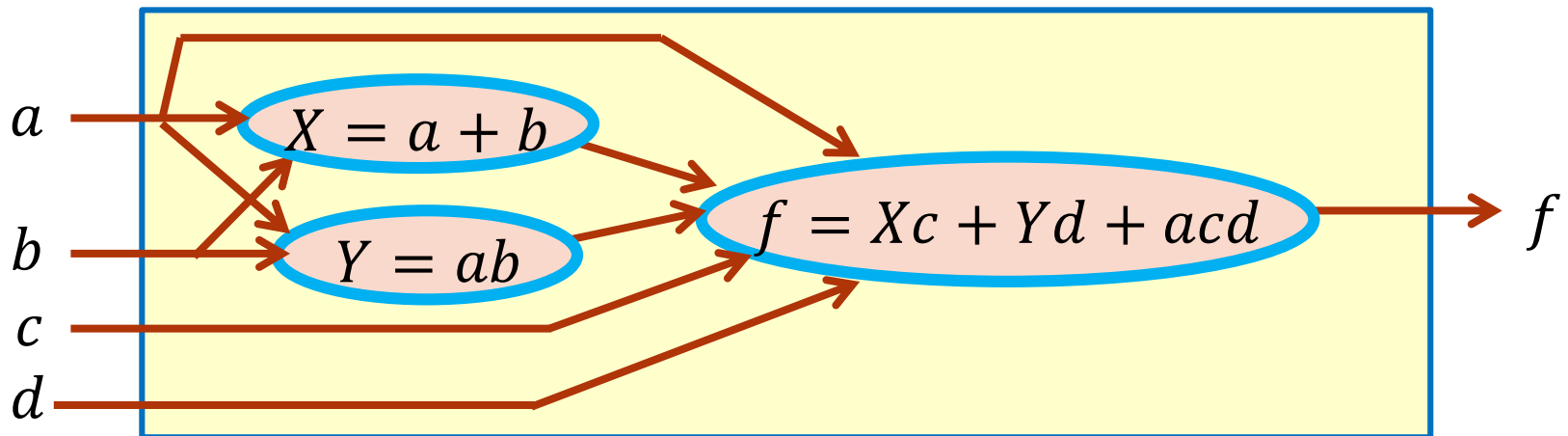
# CDCs: Why Does This Work?

$$CDC_F(X_1, \dots, X_n) = (\forall \text{ vars not used in } F) \left[ \sum_{\text{input } X_i \text{ to } F} SDC_{X_i} \right]$$

- Roughly speaking...
  - $SDC_{X_i}$ 's explain all the impossible patterns involving  $X_i$  wire input to the  $F$  node.
  - **OR** operation is just the “**union**” of all these impossible patterns involving  $X_i$ 's.
  - **Universal Quantify** removes variables **not** used by  $F$ , and does so in the right way: we want patterns that are impossible **FOR ALL** values of these removed variables.

# Compute CDCs: Example

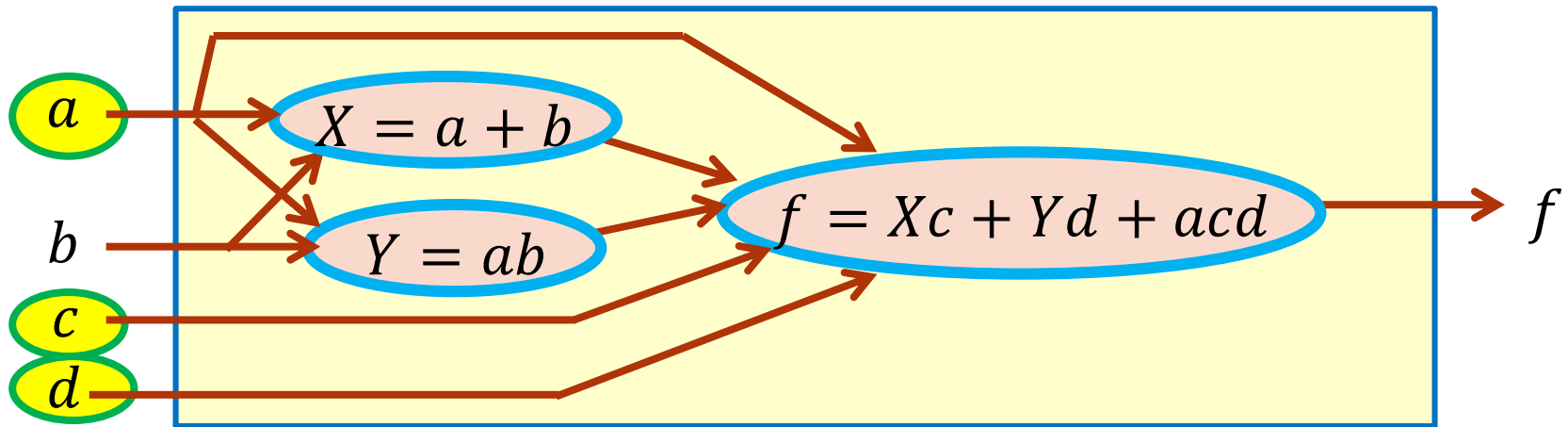
Obtain CDCs for the node  $f$



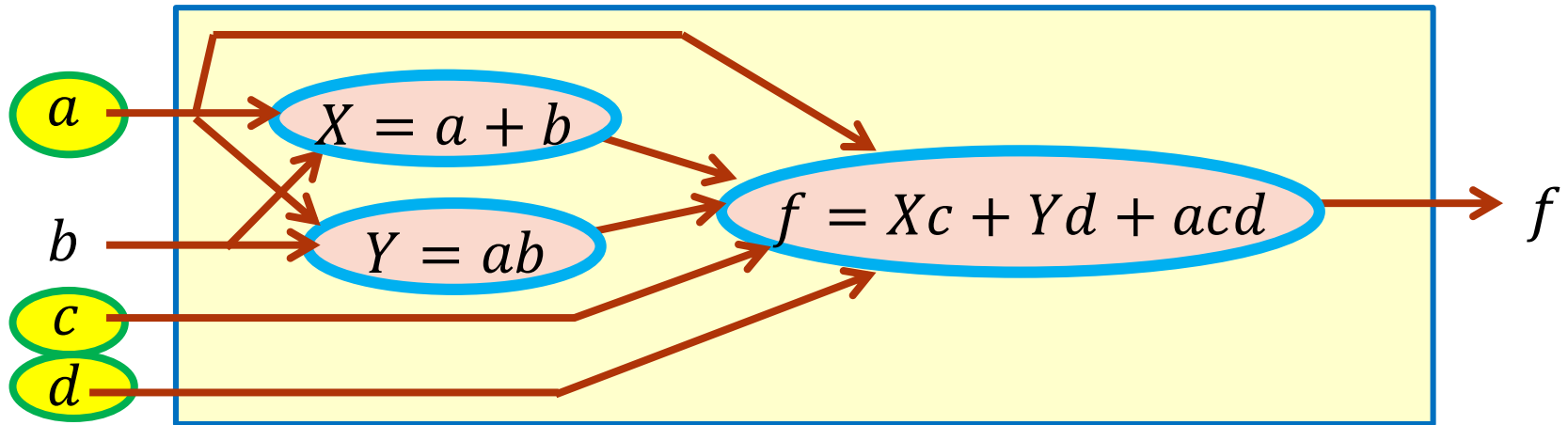
$$CDC_f(X_1, \dots, X_n) = \underbrace{(\forall \text{ vars not used in } f)}_{\text{This is } b} \left[ \underbrace{\sum_{\text{input } X_i \text{ to } f} SDC_{X_i}}_{\text{Input variables to } f \text{ are } a, c, d, X, Y} \right]$$

# Compute CDCs: Example

- What about SDCs on **primary inputs**?
  - They are just 0.
  - Why?  $SDC_a = a \oplus (\text{expression for } a) = a \oplus a = 0$ .
- **Thus**: SDCs on primary inputs have no impact on OR. We can **ignore primary inputs**.



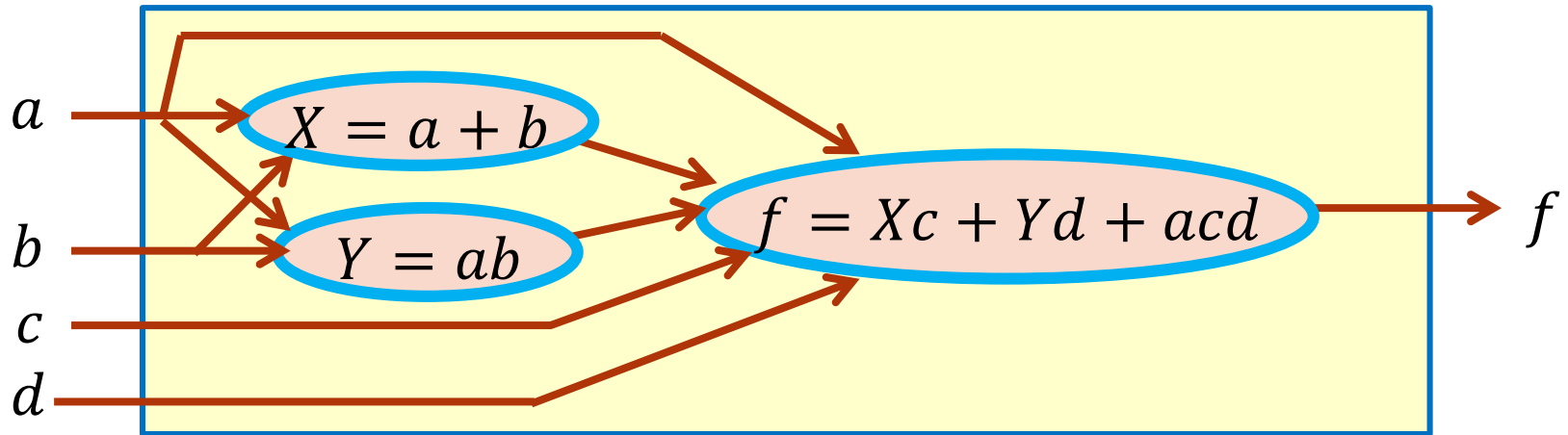
# Compute CDCs: Example



- Since we ignore primary inputs, we have ...

$$CDC_f(X_1, \dots, X_n) = \underbrace{(\forall \text{ vars not used in } f)}_{\text{This is } b} \left[ \underbrace{\sum_{\text{input } X_i \text{ to } f} SDC_{X_i}}_{\text{Only } X, Y} \right]$$

# Compute CDCs: Example

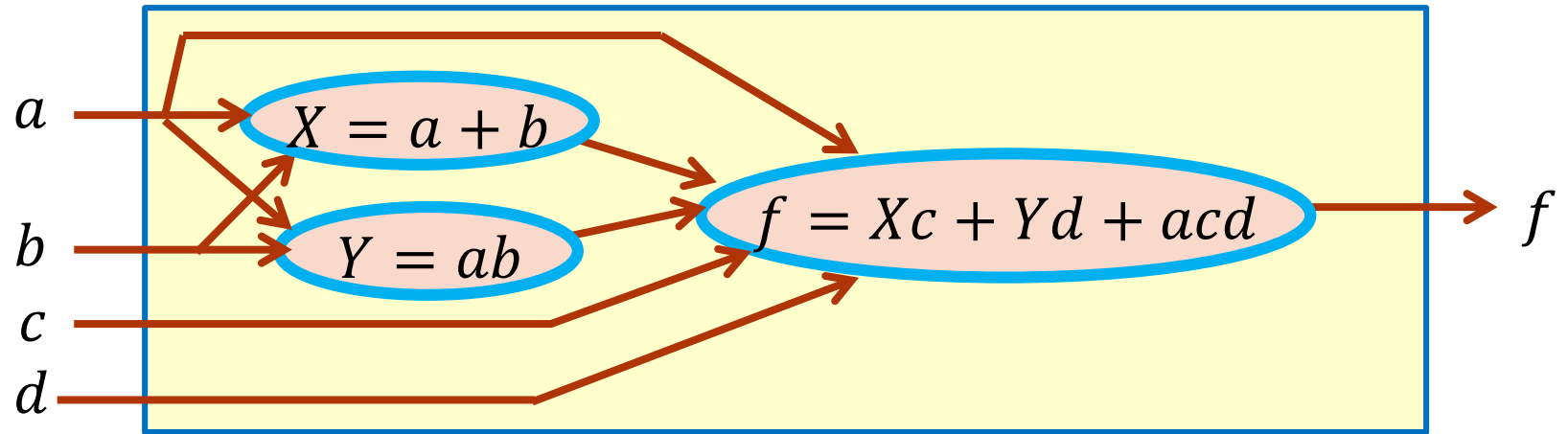


- Thus, we have:

$$\begin{aligned}
 CDC_f &= (\forall b)[SDC_X + SDC_Y] = (\forall b)[[X \oplus (a + b)] + [Y \oplus ab]] \\
 &= [[X \oplus (a + b)] + [Y \oplus ab]]_{b=1} \cdot [[X \oplus (a + b)] + [Y \oplus ab]]_{b=0} \\
 &= [\bar{X} + (Y \oplus a)] \cdot [(X \oplus a) + Y] = \bar{X}a + Y\bar{a} + \bar{X}Y
 \end{aligned}$$



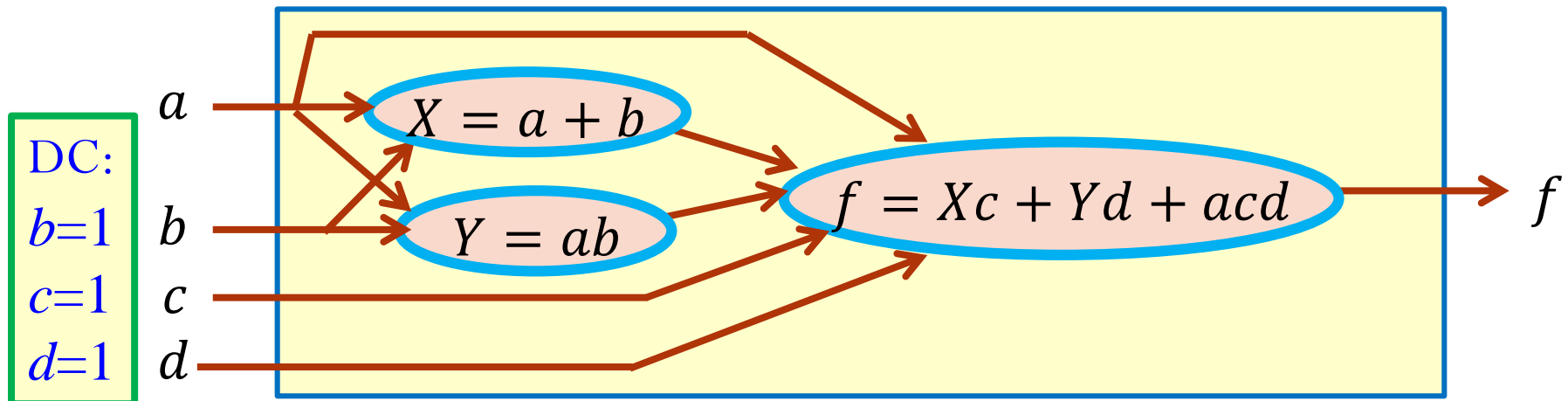
# Compute CDCs: Example



- $CDC_f = \bar{X}a + Y\bar{a} + \bar{X}Y$
- Does it **make sense**?
  - From  $CDC_f$ , **impossible patterns** are
    - $(X, a) = (0, 1)$      $a = 1 \Rightarrow X = 1$
    - $(Y, a) = (1, 0)$      $a = 0 \Rightarrow Y = 0$
    - $(X, Y) = (0, 1)$      $X = 0 \Rightarrow a = 0 \ \&\& \ b = 0 \Rightarrow Y = 0$

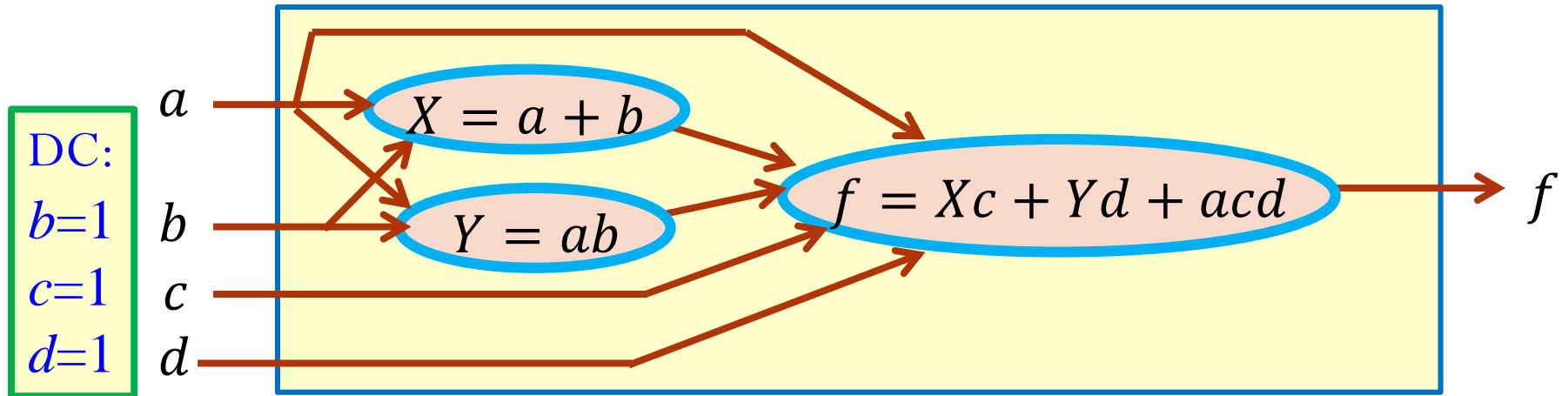
# How to Handle External CDCs?

- What if there are **external DCs** for primary inputs  $a, b, c, d$  for which we just **don't care** what  $f$  does?
  - **Answer:** Just **OR** these DCs in  $(\sum SDC_i)$  part of CDC expression.
  - Represent these DCs as a **Boolean function** that makes a 1 when the inputs are **these DCs**.



# Handling External CDCs: Example

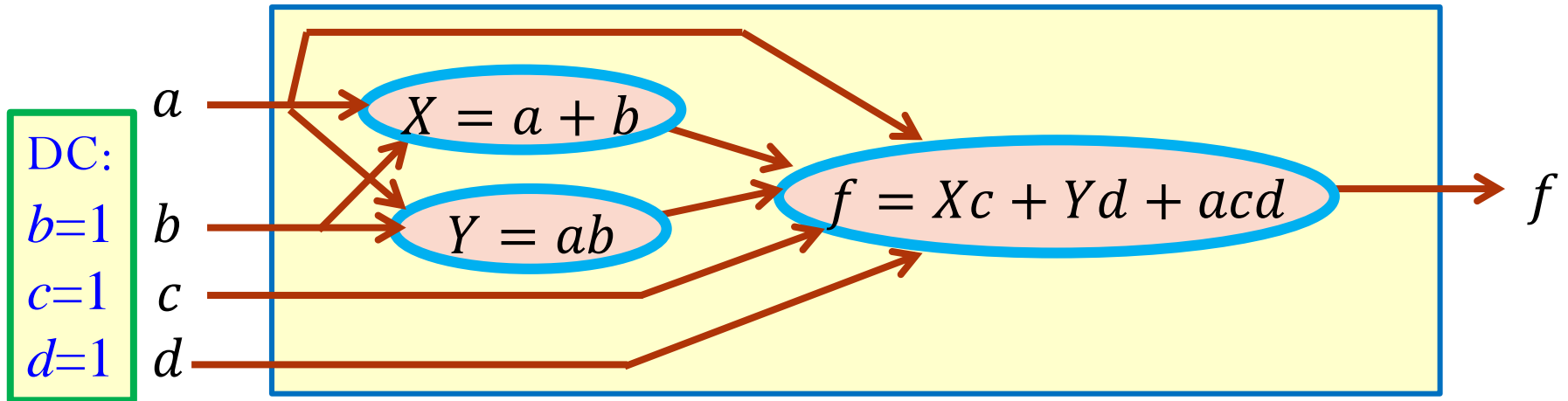
- Suppose  $(b, c, d) = (1, 1, 1)$  cannot happen.
  - How to compute  $CDC_f$  now?



$$CDC_f = (\forall b) \left[ [X \oplus (a + b)] + [Y \oplus ab] + \underbrace{bcd} \right]$$

External DCs as a **Boolean function** that makes a 1 when the pattern is **impossible**.

# Handling External CDCs: Example



$$\begin{aligned}
 CDC_f &= (\forall b) [ [X \oplus (a + b)] + [Y \oplus ab] + bcd ] \\
 &= \bar{X}a + Y\bar{a} + \bar{X}Y + \bar{a}cdX + cdY
 \end{aligned}$$

- **New impossible patterns** are

Make sense?

- $(a, c, d, X) = (0, 1, 1, 1)$       $a = 0 \ \&\& \ X = 1 \Rightarrow b = 1$

Thus,  $b = c = d = 1$

- $(c, d, Y) = (1, 1, 1)$       $Y = 1 \Rightarrow b = 1$

Thus,  $b = c = d = 1$

# CDCs: Summary

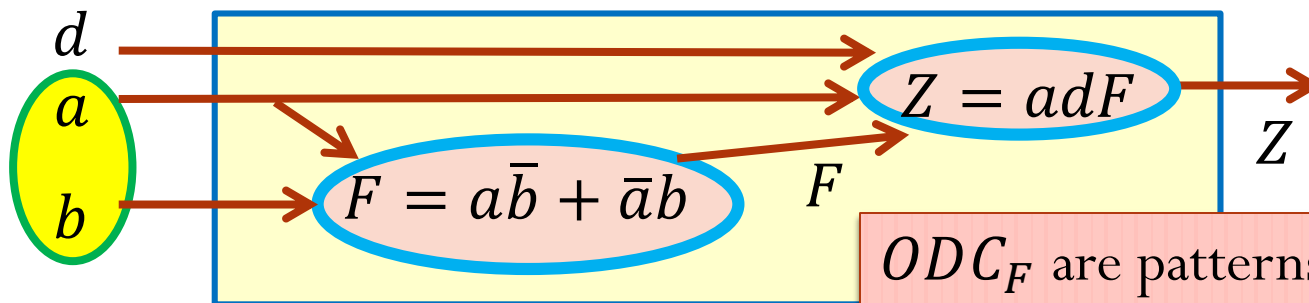
- CDCs give **impossible patterns** at input to node  $F$  – use as DCs.
  - Impossible because of the network structure of the nodes **feeding** node  $F$ .
  - CDCs can be computed mechanically from **SDCs** on wires input to  $F$ .
    - **Internal local CDCs**: computed just from SDCs on wires into  $F$ .
    - **External global CDCs**: include DC patterns at primary inputs.

# CDCs: Summary (cont.)

- But CDCs still **not all** the Don't Cares available to simplify nodes.
  - $CDC_F$  derived from the structure of nodes “**before**” node  $F$ .
  - We need to look at DCs that derive from nodes “**after**” node  $F$ .
  - These are nodes between the **output** of  $F$  and **primary outputs** of overall network.
  - These are ODCs.

# Observability Don't Cares (ODCs)

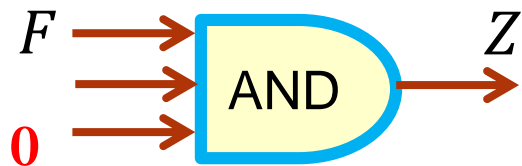
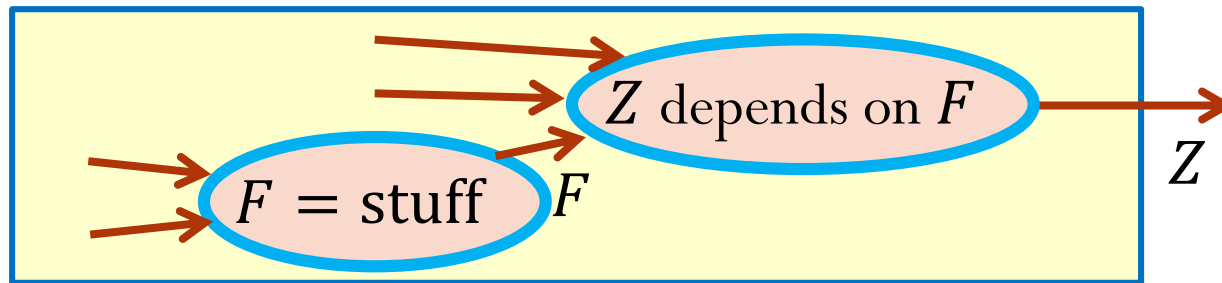
- **ODCs**: patterns that **mask** a node's output at primary output (PO) of the network.
  - So, these are **not** impossible patterns – these patterns **can occur** at node input.
  - These patterns make this node's output **not observable at primary output**.
  - “**Not observable**” for an input pattern means: Boolean value of node output **does not affect** ANY primary output.



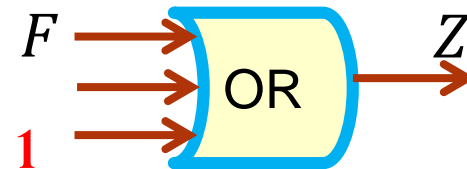
$ODC_F$  are patterns of  $(a, b)$  that make  $Z$  **insensitive** to  $F$ 's value.

# Primary Output Insensitive to F

- When is primary output  $Z$  **insensitive** to internal variable  $F$ ?
  - Means  $Z$  **independent** of value of  $F$ , given other inputs to  $Z$ .



$Z$  **insensitive** to  $F$  if  
**any other** input = 0

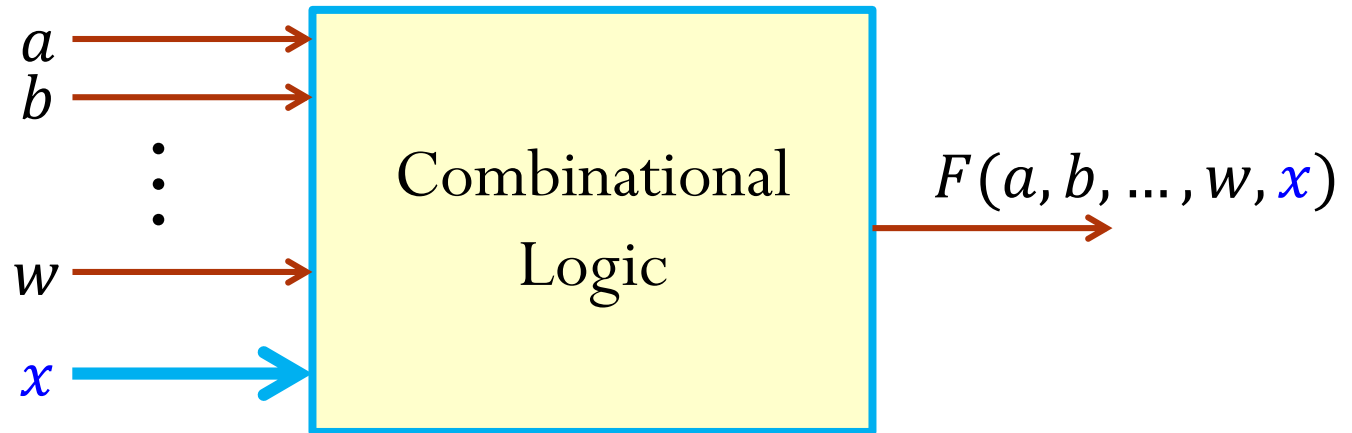


$Z$  **insensitive** to  $F$  if  
**any other** input = 1

How about the general case?



# Recall: Boolean Difference



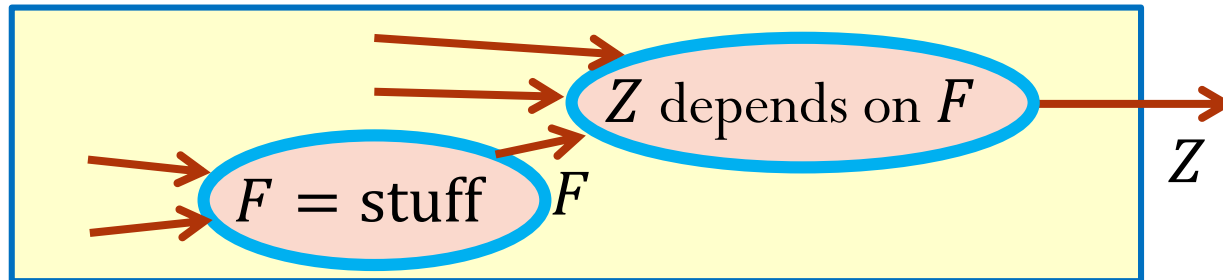
- What does **Boolean difference**  $\partial F(a, b, \dots, w, x) / \partial x = F_x \oplus F_{\bar{x}} = 1$  mean?
  - If you apply an input pattern  $(a, b, \dots, w)$  that makes  $\partial F / \partial x = 1$ , then **any change** in  $x$  will **force a change** in output  $F$ .
- What makes output  $F$  **sensitive** to input  $x$ ?
  - **Answer:** Any pattern that makes  $\frac{\partial F}{\partial x} = F_x \oplus F_{\bar{x}} = 1$ .

# Z Insensitive to F

- When is primary output  $Z$  **insensitive** to internal variable  $F$ ?
  - **Answer**: when inputs (other than  $F$ ) to  $Z$  make cofactors  $Z_F = Z_{\bar{F}}$ .
  - **Make sense**: if cofactors with respect to  $F$  are **same**,  $Z$  does not depend on  $F$ !
- How to find when cofactors are the same?
  - **Answer**: Solve for  $Z_F \oplus Z_{\bar{F}} = 1$
  - **Note**:  $Z_F \oplus Z_{\bar{F}} = 1 \Rightarrow \overline{Z_F \oplus Z_{\bar{F}}} = 1 \Rightarrow \frac{\partial Z}{\partial F} = 1$

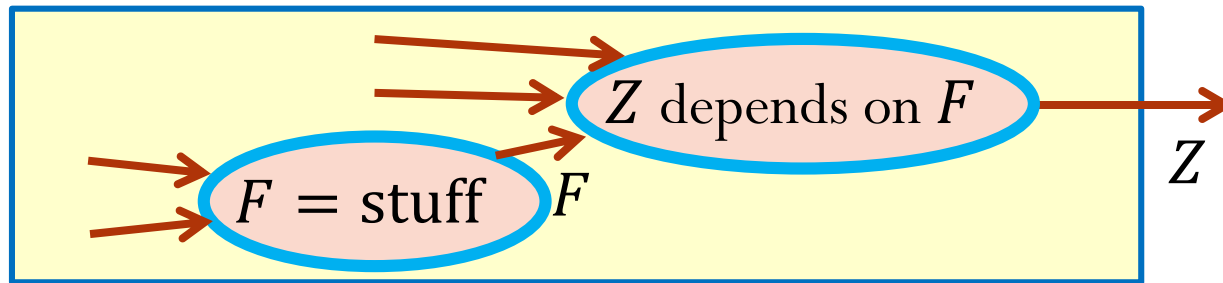
# How to Compute ODCs?

- A nice computational recipe:
  1. **Compute**  $\overline{\partial Z / \partial F}$ . Any patterns that make  $\overline{\partial Z / \partial F} = 1$  **mask** output  $F$  for  $Z$ .
  2. **Universally Quantify** away all variables that are **NOT** inputs to the  $F$  node.



$$ODC_F(X_1, \dots, X_n) = (\forall \text{ vars not used in } F) \left[ \overline{\partial Z / \partial F} \right]$$

# How to Compute ODCs?

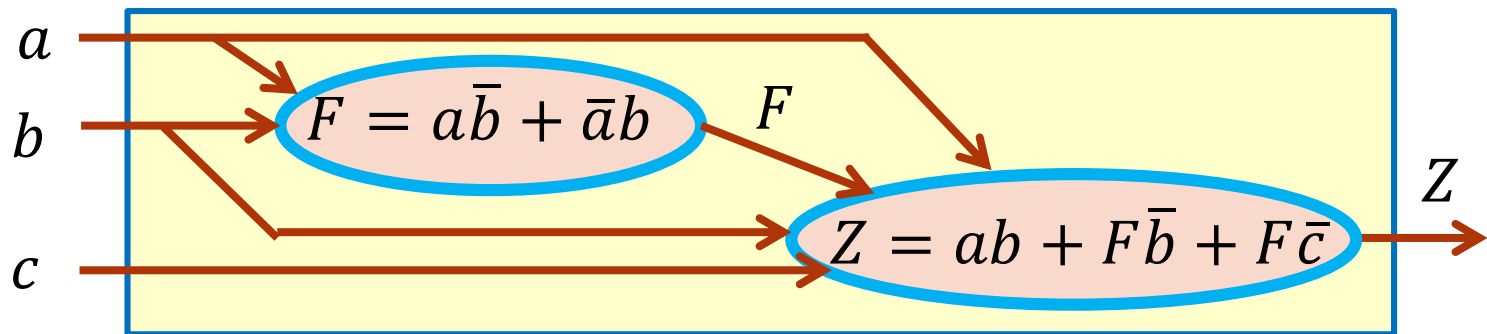


$$ODC_F(X_1, \dots, X_n) = (\forall \text{ vars not used in } F) \left[ \overline{\partial Z / \partial F} \right]$$

- **Result:** Inputs that let  $ODC_F = 1$  **mask** output  $F$  for  $Z$ , i.e., make  $Z$  **insensitive** to  $F$ .

# Compute ODCs: Example

- Obtain the ODCs for node  $F$ .

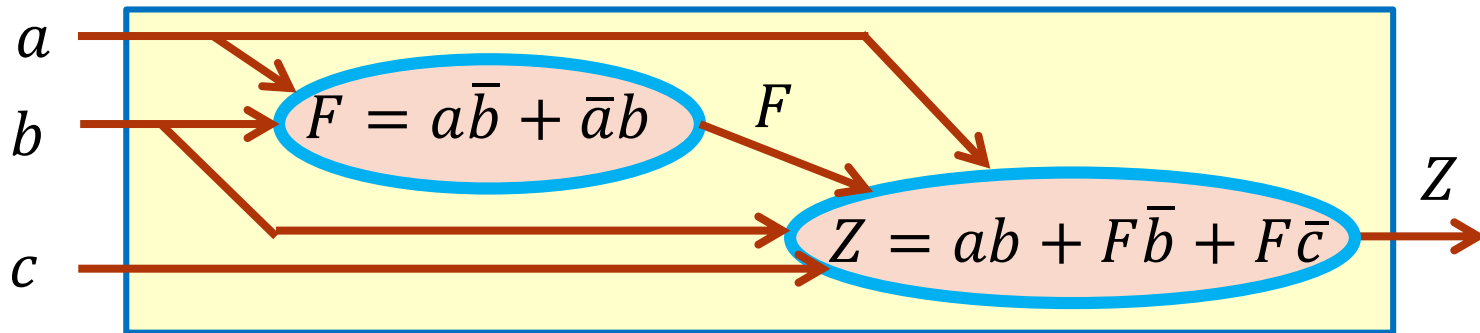


$$ODC_F(X_1, \dots, X_n) = (\forall \text{ vars not used in } F) \left[ \overline{\partial Z / \partial F} \right]$$

They are  $a, b$ 
This is  $c$

$$\begin{aligned}
 &= (\forall c) \left[ (ab + F\bar{b} + F\bar{c})_{F=1} \oplus (ab + F\bar{b} + F\bar{c})_{F=0} \right] \\
 &= (\forall c) \left[ (ab + \bar{c}) \oplus (ab) \right] = ab
 \end{aligned}$$

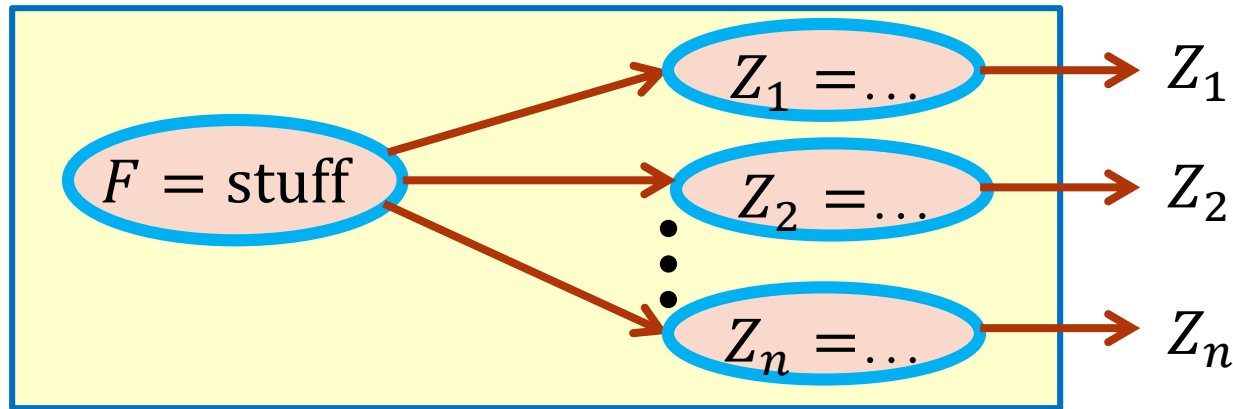
# Check: Does this ODC Make Sense?



- $ODC_F = ab$ 
  - ODC pattern is  $(a, b) = (1, 1)$
- Make sense! Because when  $(a, b) = (1, 1)$ ,  $Z = 1$  independent of  $F$ .

# ODCs: More General Case

- **Question:** what if  $F$  feeds to **many** primary outputs?
  - **Answer:** Only patterns that are **unobservable** at **ALL** outputs can be ODCs.



- Computational recipe:

$$ODC_F = (\forall \text{ vars not used in } F) \left[ \prod_{\text{Output } Z_i} \overline{\frac{\partial Z_i}{\partial F}} \right]$$

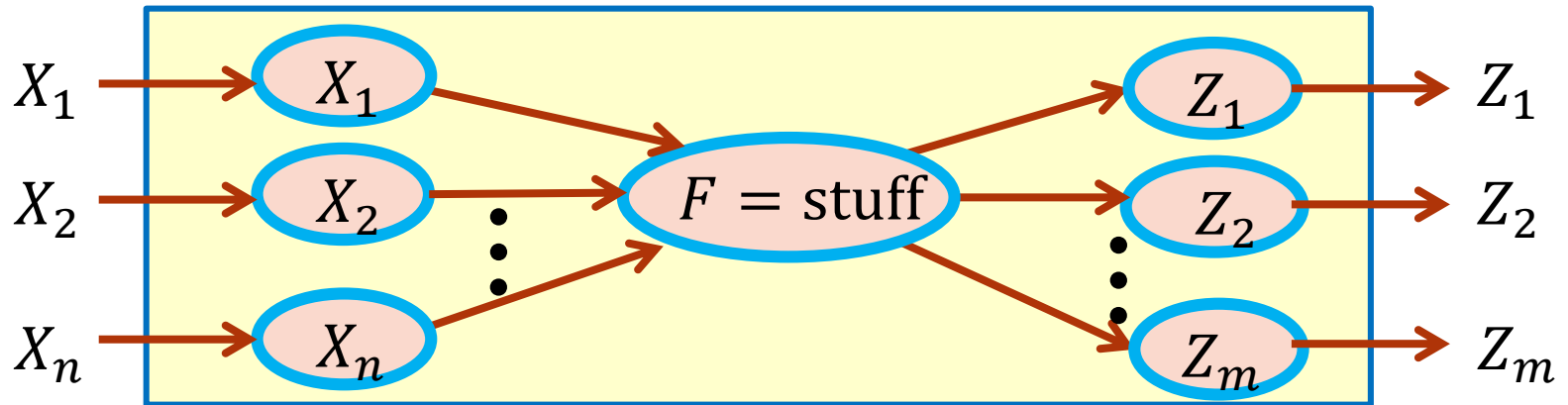
**AND** all  $n$  differences for each output  $Z_i$ .

# ODCs: Summary

- ODCs give input patterns of node  $F$  that **mask**  $F$  at **primary outputs**.
  - **Not** impossible patterns – they **can occur**.
  - Don't cares because primary output “**doesn't care**” what  $F$  is, for these patterns.
  - ODCs can be computed mechanically from  $\overline{\partial Z_i / \partial F}$  on all outputs connected to  $F$ .
- CDCs + ODCs give the “**full**” don't care set used to simplify  $F$ .
  - With these patterns, you can call something like ESPRESSO to simplify  $F$ .

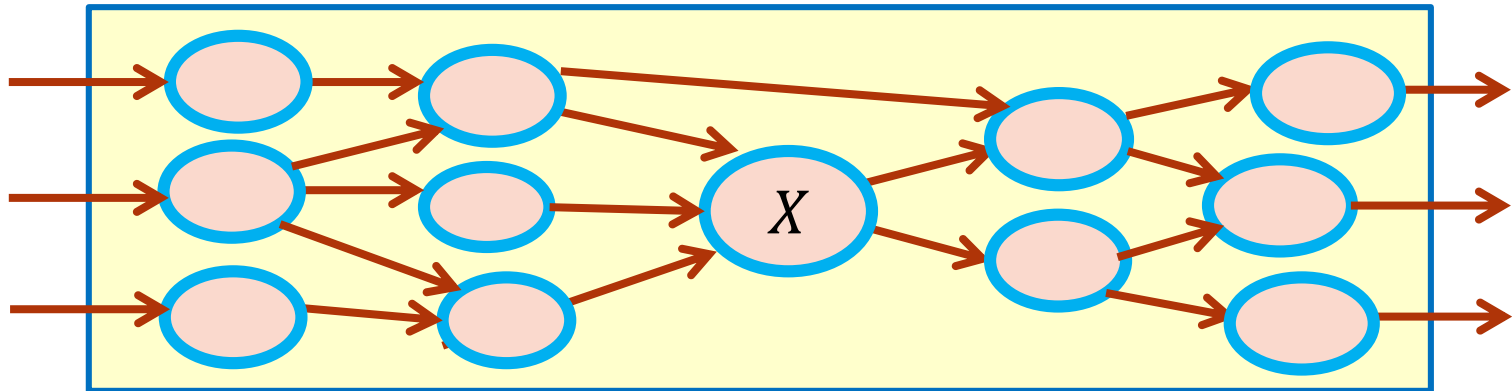


# Multi-Level Don't Cares: Are We Done?



- Yes, if your networks look **just like above**.
  - More precisely, if you only want to get CDCs from nodes **immediately** “before” you.
  - And if you only want to get ODCs for **one layer of nodes** between you and output.

# Don't Cares, In General



- But, this is what **real** multi-level logic can look like!
  - CDCs are function of **all nodes** “before”  $X$ .
  - ODCs are function of **all nodes** between  $X$  and any output.
  - In general, we can **never get all** the DCs for node  $X$  in a big network.
  - Representing all this stuff can be **explosively** large, even with BDDs

# Summary: Getting Network DCs

- How we really do it? generally **do not get all** the DCs.
  - Lots of tricks that trade off effort (time, memory) with quality (how many DCs).
  - Example: Can just extract “**local CDCs**”, which requires looking at outputs of **immediate precedent** vertices and computing from the SDC patterns, which is easy.
  - There are also **incremental**, node-by-node algorithms that walk the network to compute more of the CDC and ODC set for X, but these are more **complex**.
- For us, knowing these “limited” DC recipes is **sufficient**.