# Boolean Matching for LUT-Based Logic Blocks With Applications to Architecture Evaluation and Technology Mapping 

Jason (Jingsheng) Cong, Fellow, IEEE, and Yean-Yow Hwang


#### Abstract

In this paper, we present new Boolean matching methods for lookup table (LUT)-based programmable logic blocks (PLBs) and their applications to PLB architecture evaluations and field programmable gate array (FPGA) technology mapping. Our Boolean matching methods, which are based on functional decomposition operations, can characterize functions for complex PLBs consisting of multiple LUTs (possibly of different sizes) such as Xilinx XC4K CLBs. With these techniques, we conducted quantitative evaluation of four PLB architectures on their functional capabilities. Architecture evaluation results show that the XC4K CLB can implement $\mathbf{9 8 \%}$ of six-input and $88 \%$ of seven-input functions extracted from MCNC benchmarks, while a simplified PLB architecture is more cost effective in terms of function implementation per LUT bit. Finally, we proposed new technology mapping algorithms that integrate Boolean matching and functional decomposition operations for depth minimization. Technology mapping results show that our PLB mapping approach achieves $12 \%$ smaller depth or $15 \%$ smaller area in XC5200 FPGAs and 18\% smaller depth in XC4K FPGAs, compared to conventional LUT mapping approaches.


Index Terms-FPGA architecture, logic synthesis.

## I. Introduction

THE FIELD programmable gate array (FPGA) was introduced in the mid-1980s as an alternative for the implementation of application-specific integrated circuits (ASICs). In contrast to the cell library technology and the mask-programmable gate array technology for ASICs, an FPGA does not need to go through the fabrication process for circuit implementation and is field programmable and often field reprogrammable. Although the FPGA in general has a lower gate density and slower circuit speed, its advantages of programmability, shorter design turnaround time, and lower initial nonrecurring engineering cost (good for low to medium volume production) often offset its disadvantages. A wide range of applications has been developed using FPGAs, including fast ASIC implementation, rapid system prototyping, logic emulation, and reconfigurable computing.

[^0]FPGAs consist of three kinds of programmable elements: programmable logic blocks (PLBs), routining resources, and input-output (I/O) blocks. Each logic block contains combinational components such as multiplexers (MUXs), simple gates (e.g., AND and OR), programmable lookup tables (LUTs), and sequential components such as flip-flops. Routing resources include segmented interconnects and switching blocks. The segmented interconnects connect to the inputs and outputs of logic blocks while the switching blocks link the segments to form long routing tracks to implement routing topology. The I/O blocks can be programmed to become the primary inputs (PIs) or primary outputs (POs) of the circuits on FPGAs.

LUTs are the basic logic blocks in many FPGAs today. A $k$-input LUT ( $k$-LUT) consists of $2^{k}$ static random access memory (SRAM) cells that can store the truth table of an arbitrary $k$-input function. In many FPGAs, small LUTs are connected by fast local connections to form a PLB for implementation flexibility, better performance, and better utilization of silicon area. In contrast, some FPGAs use MUX-based logic blocks or product-term-based logic blocks. These blocks, although having more than $k$ input ports, cannot guarantee an implementation for an arbitrary $k$-input function. New universal logic module (ULM)-based FPGA logic blocks [34] had been proposed for better covering of $K$-input functions, but the coverage was still incomplete ( $99 \%$ of four-input functions using an eight-input ULM). Since LUT is widely used in today's major FPGAs and is a true ULM for functions of its input size, we focus on implementing functions using LUT-based PLBs in this paper.

A PLB can often implement one arbitrary $K$-input function, where $K$ is determined by the PLB architecture or some wide function of more than $K$ inputs. Unfortunately, it is generally a difficult problem to determine if an arbitrary given wide function can be implemented by a PLB. This is called the Boolean matching for PLB problem. Most existing technology mapping algorithms first produce a $K$-LUT mapping solution, then pack LUTs into PLBs [9], [10], [18], [19], [21], [27], [30], [31], [39]. A comprehensive survey of recent FPGA technology mapping approaches can be found in [11].

In this paper, we study Boolean matching for PLB problems. We focus on two classes of (LUT-based) PLBs: PLB1 and PLB2, as shown in Fig. 1(a) and (b), respectively. Both PLBs contain two LUTs, $F$ and $G$, at the first stage and another element $H$ at the second stage where $H$ is a 3-LUT in PLB1 and a two-input MUX in PLB2. The sets of input signals to $F$, $G$, and $H$ are denoted as $X_{F}, X_{G}$, and $X_{H}$, respectively, and

(a)

Fig. 1. Two generic PLB architectures. (a) PLB1. (b) PLB2.
the output signals of $F, G$, and $H$ are denoted as $o_{F}, o_{G}$, and $o_{H}$, respectively. For PLB1, the 3-LUT H has inputs $o_{F}, o_{G}$, and external signal $x_{H}$. For PLB2, the MUX H selects either $o_{F}$ or $o_{G}$ to output depending on the value of external selection signal $x_{H}$.

The LUT sizes $\left|X_{F}\right|$ and $\left|X_{G}\right|$ and the appearance of $x_{H}$ are architecture parameters that may vary from one PLB to another. Therefore, we shall represent PLB1 as PLB1 $\left(\left|X_{F}\right|,\left|X_{G}\right|, h\right)$, where the first two parameters are the sizes of LUTs $F$ and $G$, and $h=1$ if the $x_{H}$ line exists, otherwise $h=0$. Similarly, we shall represent PLB2 as PLB2 $\left(\left|X_{F}\right|,\left|X_{G}\right|\right)$. Under this notation, PLB1 $(4,4,1)$ is identical to the logic block of Xilinx XC4K series FPGAs [38], which is called the configurable logic block (CLB), while PLB2 $(5,5)$ is the logic block of Lucent's ORCA series FPGAs [33], which is called the programmable function unit, and $\operatorname{PLB} 2(4,4)$ is the Xilinx XC5200 CLB. When there is no confusion in the context, we refer to the two PLBs as PLB1 and PLB2.

Boolean matching for PLBs may lead to significant reduction on mapping area and circuit delay. For example, consider a six-variable function (in the Xilinx test suite used in Section VI-A), which is represented by the following sum-of-product form:

$$
\begin{aligned}
f(X)= & \bar{x}_{1} x_{3} x_{5} x_{6} \overline{\left(x_{4}+x_{2}\right)}+\bar{x}_{1} x_{3} x_{5} \bar{x}_{6}\left(x_{4}+x_{2}\right) \\
& +x_{1} \bar{x}_{2} x_{5} x_{6}\left(x_{4}+x_{2}\right)+x_{1} x_{2} \bar{x}_{3} \bar{x}_{6}\left(x_{4}+\bar{x}_{2}\right) \\
& +\bar{x}_{1} \bar{x}_{3} \bar{x}_{5} x_{6}\left(x_{4}+\bar{x}_{2}\right)+x_{1} x_{3} x_{6}\left(x_{4}+x_{2}\right) \\
& +\bar{x}_{1} \bar{x}_{3} \bar{x}_{6} \overline{\left(x_{4}+\bar{x}_{2}\right)}+x_{1} x_{5} x_{6} \overline{\left(x_{4}+\bar{x}_{2}\right)} \\
& +x_{1} \bar{x}_{6} \overline{\left(x_{4}+x_{2}\right)} .
\end{aligned}
$$

We compare three implementations of $f(X)$ produced by Chortle-crf [19], FlowMap [9], and Boolean matching algorithms, using XC4K CLBs. For the first two implementations, we applied the optimization script rugged in the sequential circuit synthesis system SIS [32], mapped the resulting gate-level network using the area-oriented mapper Chortle-crf and the depth-optimal mapper FlowMap respectively, and packed the resulting LUTs into CLBs using the efficient CLB packing procedures in [15]. The CLB networks that result from Chortle-crf and FlowMap have three levels with seven and six CLBs, respectively. However, a Boolean matching approach can obtain a single CLB implementation of $f(X)$ (Fig. 2) with each LUT implementing the following function:

$$
o_{H}=x_{6} \bar{o}_{F} o_{G}+\bar{x}_{6} o_{F} o_{G}
$$


(b)


Fig. 2. Single CLB implementation of function $f(X)$.

$$
\begin{aligned}
o_{F}= & x_{1} \bar{x}_{2} \bar{x}_{3} \bar{x}_{4}+\bar{x}_{1} x_{2} \bar{x}_{3} \bar{x}_{4}+x_{1} \bar{x}_{2} x_{3} \bar{x}_{4}+x_{1} x_{2} \bar{x}_{3} x_{4} \\
& +\bar{x}_{1} x_{3} x_{4}+\bar{x}_{1} x_{2} x_{3} \\
o_{G}= & \bar{x}_{1} \bar{x}_{3} \bar{x}_{5} x_{6}+x_{1} \bar{x}_{3} x_{5} x_{6}+\bar{x}_{3} \bar{x}_{6}+x_{3} x_{5}+x_{1} x_{3}
\end{aligned}
$$

In this implementation, $x_{1}$ and $x_{3}$ are bridged inputs of LUTs $F$ and $G$ (i.e., $x_{1}$ and $x_{3}$ are shared by $F$ and $G$ ) and $x_{6}$ is a bridged input of LUTs $H$ and $G$. Prior to this paper, it was an open problem of how to perform Boolean matching for PLBs, especially when bridged inputs are taken into consideration.

Most existing Boolean matching approaches are for ASIC design synthesis using cell libraries. A good survey can be found in [3]. Very few are targeted for LUT-based logic blocks. Boolean matching approaches were proposed in [4] and [40] for Actel's MUX-based FPGAs [1]. Their approaches cannot be applied to LUT-based PLBs directly. Mapping algorithms targeted for PLBs were proposed in [8], where linear programming was employed to compute LUT covers and PLB packings simultaneously. Functional decomposition-based mapping approaches [20], [24], [26], [37] were proposed for LUT network synthesis. None of them were targeted to implementing wide functions using PLBs. A recent work [29] studied bidecomposition of Boolean functions and applied the results to Boolean matching for some LUT-based PLBs. The results were limited. For example, the researchers were unable to solve the Boolean matching problem for the XC4K CLB completely. In this paper, we present new Boolean matching methods for PLBs. Our methods are based on classical and new functional decomposition techniques and provide a more general solution to the Boolean matching problem for LUT-based PLBs. For example, our results give exact solutions for matching functions to the XC4K CLBs. We apply our techniques to quantitative evaluation of PLB architectures (in terms of logic implementation capability) as well as to
technology mapping for FPGAs that are widely used today. Our Boolean matching approaches for PLB1 and PLB2 may be extended to other LUT-based PLBs.
This paper is organized as follows. Section II formulates the Boolean matching problem. After the introduction of classical and recent functional decomposition results in Section III, Boolean matching methods for PLB1 and PLB2 are presented in Sections IV and V, respectively. Section VI reports Boolean matching and architecture evaluation experimental results. In Section VII, we present technology mapping algorithms that employ our Boolean matching methods and report the experimental results in Section VIII. Section IX concludes the paper. Preliminary results of this study were presented in [13] and [14].

## II. Problem Formulation

Given a multilevel network of logic gates, combinational logic synthesis transforms the given network into a network of PLBs. This transformation usually includes two major steps: 1) logic optimization and 2) technology mapping. Logic optimization transforms the given network into an equivalent network that is suitable for mapping into PLBs, e.g., into a network of fewer gates and/or smaller gates. Technology mapping then transforms the resulting network into a PLB network of minimal cost, where the cost could be network area of delay. Conventional technology mapping usually include covering the gate-level network with LUTs and packing LUTs into PLBs. The FlowMap and the Chortle-crf algorithms are two widely used technology mapping algorithms. However, as shown in the previous section, they could generate suboptimal solutions compared to Boolean matching approaches. In this paper, we focus on the Boolean matching for PLBs and then apply our results to technology mapping.

For each LUT-based PLB architecture, we define the characteristic number $K$ of the PLB to be the largest number such that any function of $K$ or fewer variables is realizable by the PLB. Functions of more than $K$ variables are called wide functions with respect to the PLB. Clearly, if the PLB has more than $K$ inputs, it can implement some wide functions. For example, if $\left|X_{F}\right|=\left|X_{G}\right|=4$ in PLB1 and PLB2, then either PLB can implement any function of up to five variables $(K=5)$ or some wide function of up to nine variables. The Boolean matching problem for PLB is to determine, for a given wide function with respect to the PLB, whether the function can be implemented by the PLB.

Boolean Matching for PLB: Given a wide function $f(X)$ with respect to a PLB $A$, determine if $f(X)$ can be realized by a single PLB $A$.

In general, Boolean matching considers input negation and/or permutation, output inversion, bridging of inputs, and constant assignments to some inputs. For LUT-based PLBs, however, input negation, input permutation and bridging in one LUT, output inversion, and constant assignments to unused inputs do not affect the matching feasibility. Therefore, only input partitioning and input sharing among LUTs are relevant factors to our Boolean matching for PLB problem.

## III. Functional Decomposition

Our solution to the Boolean matching for PLB problem relies on functionally decomposing wide functions. In this section, we shall introduce decomposition forms that are closely related to PLB matching and review existing results and present new results on functional decomposition.

## A. Preliminaries

Let $f(X)$ denote Boolean function $f\left(x_{1}, x_{2}, \ldots, x_{n}\right)$, where $X=\left\{x_{1}, x_{2}, \ldots, x_{n}\right\}$. If PLB1 implements $f(X)$, we can represent $f(X)$ as $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$, where $X_{F} \cup X_{G} \cup\left\{x_{H}\right\}=X$. This implies the existence of a functional decomposition of $f(X)$. If PLB2 implements $f(X)$, we can represent $f(X)$ as $\bar{x}_{H} \cdot y_{1}\left(X_{F}\right)+x_{H} \cdot y_{2}\left(X_{G}\right)$. This is the Shannon expansion of $f(X)$ with respect to the variable $x_{H}$. Therefore, decomposition of functions plays an important role in our Boolean matching approaches. Before presenting our approaches, we will give a brief review of classical and recent functional decomposition results.

Given a Boolean function $f(X)$, let $f_{\bar{x}_{i}}$ and $f_{x_{i}}$ denote the cofactors of $f(X)$ with respect to variable $x_{i}$. Cofactors of $f(X)$ with respect to multiple variables are defined in a similar way. For example, $f_{x_{1} \bar{x}_{2}}=\left.f(X)\right|_{x_{1}=1, x_{2}=0}$. The Shannon expansion of $f(X)$ with respect to $x_{i}$ is $f(X)=x_{i} f_{x_{i}}+\bar{x}_{i} f_{\bar{x}_{i}}$. The Shannon expansion of $f(X)$ with respect to multiple variables is defined in a similar way. The cofactor set of $f(X)$ with respect to a set $B \subseteq X$, denoted $\operatorname{cs}_{B}(f)$, is the set of all distinct cofactors of $f(X)$ with respect to the variables in $B$. We denote the support of function $f(X)$ as $\sup (f)=X$.

Given a function $f(X)$ and a set $B=\left\{x_{1}, x_{2}, \ldots, x_{b}\right\} \subset X$, the disjoint functional decomposition of $f(X)$ under $B$ represents $f(X)$ in the following form:

$$
f(X)=g\left(y_{1}(B), y_{2}(B), \ldots, y_{t}(B), x_{b+1}, \ldots, x_{n}\right)
$$

We call the set $B$ the bound set and the set $X-B$ of remaining variables the free set of the decomposition. The decomposition is nontrivial if $t<b$. If $t=1$, it is called a simple disjoint decomposition. Each $y_{i}(B)$ is called an encoding function. Let $Y=\left\{y_{1}, y_{2}, \ldots, y_{t}\right\}$ and $Y(B)=\left\{y_{1}(B), y_{2}(B), \ldots, y_{t}(B)\right\}$. We can write the decomposition as $f(X)=g(Y(B), X-B)$.

Two other functional decomposition forms are closely related to the disjoint decomposition: the nondisjoint decomposition and the partially dependent decomposition. A nondisjoint decomposition of $f(X)$ is the case when some bound set variables appear in the support of $g$. Let $S=\left\{x_{j}, \ldots, x_{b}\right\} \subset B$ for some $j>1$. Then, a nondisjoint functional decomposition of $f(X)$ is to represent $f(X)$ as $g(Y(B), S, X-B)=f(X)$. The decomposition is nontrivial when $|Y|+|S|<|B|$. Variables in $S$ are called nondisjoint variables. When $S=\emptyset$, the decomposition becomes disjoint.

A partially dependent decomposition is the case when the support of some encoding function is a strict subset of $B$, e.g., $\sup \left(y_{1}\right)=B_{m}$, where $B_{m}=\left\{x_{1}, \ldots, x_{m}\right\} \subset B$. Such an encoding function is called a partially dependent encoding function. If the support of some partially dependent encoding function contains only one variable, this encoding function can be
replaced by the variable; hence, the decomposition becomes a nondisjoint decomposition.

If $f(X)$ can be represented in the form $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right)\right)$, where $X_{1} \cup X_{2}=X$, it is called a bidecomposition of $f(X)$. A bidecomposition is disjoint if $X_{1} \cap X_{2}=\emptyset$, otherwise, it is nondisjoint. Bidecomposition plays an important role in the Boolean matching for PLB1 when the $x_{H}$ line is not used or when there are bridged inputs.

For any implementation of $f(X)$ on PLB1, existence of bridged inputs implies nondisjoint functional decomposition or bidecomposition of $f(X)$ or a combination of them. For PLB2, existence of bridged inputs implies the two cofactors share common variables.

## B. Existence Conditions

We now briefly review the existence conditions for various forms of functional decomposition. Ashenhurst [2] gave the existence condition for simple disjoint decomposition.

Theorem 1: There exists a simple disjoint decomposition $g\left(y_{1}(B), X-B\right)$ of $f(X)$ under the bound set $B$ if and only if $\left|\operatorname{cs}_{B}(f)\right| \leq 2$ [2].

The condition was extended to general disjoint decomposition $(t \geq 1)$ by Curtis [17].

Theorem 2: There exists a disjoint decomposition $g(Y(B), X-B)$ of $f(X)$ under the bound set $B$ if and only if $\left|\operatorname{cs}_{B}(f)\right| \leq 2^{|Y|}$ [17].

The existence of disjoint bidecomposition $g\left(y_{1}\left(X_{1}\right), y_{2}\left(X_{2}\right)\right)$ of $f(X)$ can be determined by verifying the existence of simple disjoint decomposition of $f(X)$ under the bound sets $X_{1}$ and $X_{2}$, respectively. In other words, a disjoint bidecomposition can be obtained by combining two simple disjoint decompositions. This result is implied by the following theorem.

Theorem 3: Let $X_{1}, X_{2}, \ldots, X_{t}$ be a disjoint partition of $X$. A functional decomposition $g\left(y_{1}\left(X_{1}\right), y_{2}\left(X_{2}\right), \ldots, y_{t}\left(X_{t}\right)\right)$ of $f(X)$ exists if and only if there exists simple disjoint decomposition of $f(X)$ under each bound set $X_{1}, X_{2}, \ldots, X_{t}$, respectively [2].

For nondisjoint bidecompositions of $f(X)$, it was shown in [29] that they can be obtained by applying disjoint bidecomposition to the cofactors of $f(X)$. In fact, a few forms of nondisjoint decomposition that correspond to different PLB input bridging patterns can be obtained in a similar way. Because of this, we shall defer to the next section to present the existence condition when we introduce the Boolean matching approach for each PLB input bridging pattern.

A few approaches for partially dependent decomposition and nondisjoint decomposition were proposed in the past few years [13], [20], [25], [26], [28], with different perspectives and algorithmic characteristics. In particular, an existence condition for both partially dependent decomposition and nondisjoint decomposition was given in [13]. The condition can be used to compute one partially dependent encoding function or nondisjoint variable efficiently.

Theorem 4: Let $B_{m} \subset B$. There exists a partially dependent decomposition $g(Y(B), X-B)$ of $f(X)$ under the bound set $B$
with $\sup \left(y_{1}\right)=B_{m}$ if and only if $\operatorname{cs}_{B_{m}}(f)$ can be partitioned into two sets $\mathrm{CS}_{0}$ and $\mathrm{CS}_{1}$ such that [13]

$$
\begin{aligned}
& \left|\bigcup_{h \text { incs }} \operatorname{cs}_{B-B_{m}}(h)\right| \leq 2^{|Y|-1} \quad \text { and } \\
& \left|\bigcup_{h \mathrm{incs}_{1}} \operatorname{cs}_{B-B_{m}}(h)\right| \leq 2^{|Y|-1}
\end{aligned}
$$

Proof-Only If: Assume $f(X)=g(Y(B), X-B)$ and $\sup \left(y_{1}\right)=B_{m}$. Let $p$ be a minterm in the Boolean space defined on the variable set $B_{m}$. Define the set $\mathrm{cs}_{0}=\{f(p, X-$ $\left.\left.B_{m}\right) \mid y_{1}(p)=0\right\}$. Let $\operatorname{cs}_{0, B}(g)$ denote the set of cofactors $g\left(0, y_{2}(B), \ldots, y_{|Y|}(B), X-B\right)$ with respect to variables in $B$. First, the set $\cup_{h \in \operatorname{cs}_{0}} \operatorname{cs}_{B-B_{m}}(h)$ of cofactors is a subset of $\operatorname{cs}_{0, B}(g)$ because every function $h \in \operatorname{cs}_{0}$ has the form

$$
\begin{aligned}
& h=\left.f\left(p, X-B_{m}\right)\right|_{y_{1}(p)=0}=g\left(0, y_{2}\left(p, B-B_{m}\right), \ldots,\right. \\
& \left.\quad y_{|Y|}\left(p, B-B_{m}\right), X-B\right)\left.\right|_{y_{1}(p)=0}
\end{aligned}
$$

Second, $\left|\operatorname{Cs}_{0, B}(g)\right| \leq 2^{|Y|-1}$ because $\operatorname{cs}_{0, B}(g)$ must be a subset of $\operatorname{cs} \boldsymbol{S}_{Y}\left(g_{\bar{y}_{1}}\right)$, the set of cofactors $g\left(0, y_{2}, \ldots, y_{|Y|}, X-\right.$ $B)$ with respect to variables $y_{2}, \ldots, y_{m}$. Therefore, $\left|\cup_{h \in \operatorname{cs}_{0}} \operatorname{cs}_{B-B_{m}}(h)\right| \leq 2^{|Y|-1}$. Similarly, define $\operatorname{CS}_{1}=\left\{f\left(p, X-B_{m}\right) \mid y_{1}(p)=1\right\}$ and we have $\left|\cup_{h \in \operatorname{cs}_{1}} \operatorname{cs}_{B-B_{m}}(h)\right| \leq 2^{|Y|-1}$. If $\operatorname{cs}_{0}$ and $\operatorname{cs}_{1}$ form a partition of $\operatorname{cs}_{B_{m}}(f)$, we have proved the condition. Otherwise, consider $\operatorname{cs}_{0}^{\prime}=\operatorname{cs}_{0}-\operatorname{cs}_{1}$. Then, $\mathrm{cs}_{0}^{\prime}$ and $\mathrm{cs}_{1}$ form a partition that satisfies the necessary condition.

If: Assume $\cup_{h \in \operatorname{cs}_{0}} \operatorname{cs}_{B-B_{m}}(h)$ and $\cup_{h \in \operatorname{cs}_{1}} \operatorname{cs}_{B-B_{m}}(h)$ each contains at most $2^{t-1}$ members. Let $p$ be a minterm in the Boolean space defined on the variable set $B_{m}$. Then, $f\left(p, X-B_{m}\right)$ represents a cofactor in $\operatorname{cs}_{B_{m}}(f)$. Define Boolean function $y_{1}\left(B_{m}\right)$ to be $y_{1}(p)=0$ if $f\left(p, X-B_{m}\right) \in \operatorname{cs}_{0}$ and $y_{1}(p)=1$ if $f\left(p, X-B_{m}\right) \in \operatorname{CS}_{1}$. Let $f_{-}\left(X-B_{m}\right)$ be an arbitrary cofactor in $\operatorname{cs}_{0}$. We define $f_{0}(X)$ to be $f_{0}\left(p, X-B_{m}\right)=f\left(p, X-B_{m}\right)$ if $y_{1}(p)=0$ and $f_{0}\left(p, X-B_{m}\right)=f_{-}\left(X-B_{m}\right)$ if $y_{1}(p)=1$ for every minterm $p$. Then, the cofactor set of $f_{0}(X)$ under the bound set $B, \operatorname{cs}_{B}\left(f_{0}\right)$ is identical to $\cup_{h \in \operatorname{cs}_{0}} \operatorname{cs}_{B-B_{m}}(h)$, which contains at most $2^{t-1}$ members.

According to Theorem 2, there exists a disjoint decomposition of $f_{0}(X)$ under the bound set $B$ with $t-1$ encoding functions: $f_{0}(X)=g_{0}\left(w_{2}(B), \ldots, w_{t}(B), X-B\right)$. Similarly, let $f_{+}(X-$ $B_{m}$ ) be an arbitrary cofactor in $\mathrm{CS}_{1}$. Define function $f_{1}(X)$ to be $f_{1}\left(p, X-B_{m}\right)=f_{+}\left(X-B_{m}\right)$ if $y_{1}(p)=0$ and $f_{1}(p, X-$ $\left.B_{m}\right)=f\left(p, X-B_{m}\right)$ if $y_{1}(p)=1$ for every minterm $p$. Then, there exists a disjoint decomposition of $f_{1}(X)$ with $t-1$ encoding functions: $f_{1}(X)=g_{1}\left(z_{2}(B), \ldots, z_{t}(B), X-B\right)$. Because $f=\bar{y}_{1} f_{0}+y_{1} f_{1}$, we have $f=\bar{y}_{1} g_{0}+y_{1} g_{1}$. Define $y_{i}=$ $\bar{y}_{1} w_{i}+y_{1} z_{i}(2 \leq i \leq t)$ and $g=\bar{y}_{1} g_{0}+y_{1} g_{1}$. Then, we have

$$
\begin{aligned}
f(X)= & \bar{y}_{1}\left(B_{m}\right) \cdot g_{0}\left(y_{2}(B), \ldots, y_{t}(B), X-B\right) \\
& +y_{1}\left(B_{m}\right) \cdot g_{1}\left(y_{2}(B), \ldots, y_{t}(B), X-B\right) \\
= & g\left(y_{1}\left(B_{m}\right), y_{2}(B), \ldots, y_{m}(B), X-B\right) .
\end{aligned}
$$

This proves the sufficient condition.

For finding multiple partially dependent encoding functions, the authors in [13] took an approach similar to that in [36] and [37]. The existence of nondisjoint decomposition of $f(X)$ with the nondisjoint variable $x_{i}$ can be checked by setting $B_{m}=$ $\left\{x_{i}\right\}$ in Theorem 4. For the general case of multiple nondisjoint variables, we have the following theorem.
Theorem 5 : Given a bound set $B$ and a set of nondisjoint variables $S \subset B$, there exists a nondisjoint decomposition $g(Y(B), S, X-B)$ of $f(X)$ if and only if $\left|\operatorname{cs}_{B-S}(h)\right| \leq 2^{|Y|}$ for every cofactor $h \in \operatorname{CS}_{S}(f)$ [13].

Proof: We prove it by mathematical induction on $|S|$.
Basis: This theorem holds for $|S|=1$ according to Theorem 4 with $B_{m}=S$. Without loss of generality, assume $S=$ $\left\{x_{1}\right\}$. Then, $\operatorname{cs}_{B_{m}}(f)=\left\{f_{\bar{x}_{1}}, f_{x_{1}}\right\}$. The only possible partition is $\mathrm{CS}_{0}=\left\{f_{\bar{x}_{1}}\right\}$ and $\operatorname{CS}_{1}=\left\{f_{x_{1}}\right\}$. Theorem 4 then implies the basis.

Hypothesis: Assume this theorem holds for $|S|=k-1$. Now, we prove the case for $|S|=k$.

Only If: Without loss of generality, assume $x_{1} \in S$. Define $S^{\prime}=S-\left\{x_{1}\right\}$. Then, $f_{\bar{x}_{1}}=g_{\bar{x}_{1}}\left(Y(B), S^{\prime}, X-B\right)$, a decomposition with $k-1$ nondisjoint variables. According to induction hypothesis, every cofactor $h \in \operatorname{cs}_{S^{\prime}}\left(f_{\bar{x}_{1}}\right)$ satisfies $\left|\operatorname{cs}_{B-S}(h)\right| \leq 2^{|Y|}$. Similarly, every cofactor $h \in \operatorname{cs}_{S^{\prime}}\left(f_{x_{1}}\right)$ satisfies $\left|\operatorname{cs}_{B-S}(h)\right| \leq 2^{|Y|}$. However, $\operatorname{cs}_{S}(f)=\operatorname{cs}_{S^{\prime}}\left(f_{\bar{x}_{1}}\right) \cup$ $\operatorname{cs}_{S^{\prime}}\left(f_{x_{1}}\right)$. Therefore, $\left|\operatorname{cs}_{B-S}(h)\right| \leq 2^{|Y|}$ for every $h \in \operatorname{cs}_{S}(f)$.

If: Assume $\left|\operatorname{cs}_{B-S}(h)\right| \leq 2^{t}$ for every $h \in \operatorname{cs}_{S}(f)$. Define $B^{\prime}=B-\left\{x_{1}\right\}$. Since $\operatorname{cs}_{S}(f)=\operatorname{cs}_{S^{\prime}}\left(f_{\bar{x}_{1}}\right) \cup \operatorname{cs}_{S^{\prime}}\left(f_{x_{1}}\right)$, every cofactor $h \in \operatorname{cs}_{S^{\prime}}\left(f_{\bar{x}_{1}}\right)$ satisfies $\left|\operatorname{cs}_{B-S}(h)\right| \leq 2^{t}$. According to induction hypothesis, $f_{\bar{x}_{1}}=g_{0}\left(W\left(B^{\prime}\right), S^{\prime}, X-B\right)$, where $|W|=t$. Similarly, $f_{x_{1}}=g_{1}\left(Z\left(B^{\prime}\right), S^{\prime}, X-B\right)$ and $|Z|=t$. Define $g=\bar{x}_{1} g_{0}+x_{1} g_{1}$ and $y_{i}=\bar{x}_{1} w_{i}+x_{1} z_{i}$, where $w_{i}\left(B^{\prime}\right) \in W\left(B^{\prime}\right)$ and $z_{i}\left(B^{\prime}\right) \in Z\left(B^{\prime}\right)$. Then, we have

$$
\begin{aligned}
f(X)= & \bar{x}_{1} \cdot g_{0}\left(Y(B), S^{\prime}, X-B\right) \\
& +x_{1} \cdot g_{1}\left(Y(B), S^{\prime}, X-B\right) \\
= & g(Y(B), S, X-B)
\end{aligned}
$$

This proves the theorem.

## IV. Boolean Matching Approaches for PLB1

In this section, we consider matching wide functions to PLB1 in Configurations A, B, C, and D shown in Fig. 3 obtained from different bridging status of input $x_{H}$. In Configuration A, $x_{H}$ does not feed to LUT $H$; in Configuration $\mathrm{B}, x_{H}$ feeds to LUT $H$ only; in Configuration C, $x_{H}$ feeds to all three LUTs, while in Configuration D, $x_{H}$ feeds to LUTs $H$ and $G$. For each configuration, there might be bridged inputs to LUTs $F$ and $G$, shown as dash lines in Fig. 3. There can be multiple bridged inputs to $F$ and $G$ as long as the total number of distinct inputs matches that of the wide function. For example, up to two bridged inputs to $F$ and $G$ are allowed when we consider matching a six-variable function to a XC4K CLB in Configuration D. It should be clear that the four configurations exhaust all possible ways of using PLB1. Hence, the Boolean matching problem for PLB1 can be solved by matching functions to the four configurations individually.


Fig. 3. Four PLB1 configurations. (a) Configuration A. (b) Configuration B. (c) Configuration C. (d) Configuration D.

## A. PLB1 in Configuration A

If PLB1 in Configuration A implements $f(X)$, a bidecomposition $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right)\right)$ must exist for $f(X)$. If bridged inputs exist in the implementation, the decomposition is nondisjoint. Otherwise, it is disjoint. For the case without bridged inputs, an implementation, if it exists, can be obtained by combining simple disjoint decompositions of $f(X)$ under the bound sets $X_{F}$ and $X_{G}$, respectively (according to Theorem 3).

We now consider the case when PLB1 implements $f(X)$ with bridged inputs. Let $X_{F}=X_{1} \cup X_{3}$ and $X_{G}=X_{2} \cup X_{3}$, where $X_{3}$ is the set of bridged inputs. In general, $\left|X_{3}\right|>1$. For the case where input $x_{i}$ is bridged, we have the following theorem.

Theorem 6 : Let $X_{F}=X_{1} \cup\left\{x_{i}\right\}, X_{G}=X_{2} \cup\left\{x_{i}\right\}$. A nondisjoint bidecomposition $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right)\right)$ of $f(X)$ exists if and only if there exist bidecompositions $f_{\bar{x}_{i}}=g_{0}\left(y_{01}\left(X_{1}\right), y_{02}\left(X_{2}\right)\right)$ and $f_{x_{i}}=g_{1}\left(y_{11}\left(X_{1}\right), y_{12}\left(X_{2}\right)\right)$ such that $g_{0}\left(y_{1}, y_{2}\right)=g_{1}\left(y_{1}, y_{2}\right)$ [29].

To determine if a function $f(X)$ can be implemented with PLB1 in Configuration A, we first enumerate bipartitions of $X$ and test $f(X)$ for disjoint bidecompositions based on Theorem 3. If none can be found, we choose variables in $X$ as bridged inputs enumeratively and test $f(X)$ for nondisjoint bidecompositions based on Theorem 6. When multiple bridged inputs are explored, we decompose cofactors recursively. For example, to test for a matching with bridged inputs $x_{i}$ and $x_{j}$, we compute all nondisjoint bidecompositions for cofactors $f_{\bar{x}_{i}}$ and $f_{x_{i}}$ with $x_{j}$ being the nondisjoint variable and then determine if the resulting bidecompositions can satisfy the condition in Theorem 6 with $x_{i}$ being the nondisjoint variable. If the condition can be satisfied, then bidecompositions of $f(X)$ with two bridged inputs $x_{i}$ and $x_{j}$ exist.

Note that the bidecompositions of cofactors $f_{\bar{x}_{i}}$ and $f_{x_{i}}$ are not unique in general. When we compute nondisjoint bidecompositions, in order to check if $g_{0}$ can be equal to $g_{1}$ for some feasible bidecompositions, we invert $y_{01}\left(X_{1}\right), y_{02}\left(X_{2}\right)$, or both and compute the corresponding $g_{0}$ function under each case. In total, we obtain four equivalent bidecompositions of $f_{\bar{x}_{i}}$. Then,
we compare each resulting function $g_{0}$ with $g_{1}$ for a match. Since $\left|\sup \left(g_{0}\right)\right|=2$, we have exhausted all bidecompositions of $f_{\bar{x}_{i}}$.

A special case that requires particular attention is when one of the cofactors is a constant (zero or one). For example, assume that $f_{\bar{x}_{i}}$ is a constant. Then, we can always obtain a nondisjoint bidecomposition of $f(X)$ by duplicating $g_{1}$ for $g_{0}$ followed by producing constant encoding functions $y_{01}\left(X_{1}\right)$ and $y_{02}\left(X_{2}\right)$ such that $g_{0}$ will return the same value as $f_{\bar{x}_{i}}$. The condition in Theorem 6 is, thus, satisfied for this special case. When both cofactors are constant, we choose an arbitrary two-input function, e.g., the XOR function, for both $g_{0}$ and $g_{1}$ function and choose (constant) encoding functions accordingly.

We illustrate the special case by an example. Let $X=\{a, b, c, d, e, j\}$ and $f(X)=\bar{a}+a(b c d+e j)$. We want to check if a nondisjoint bidecomposition $f(X)=g\left(y_{1}(a, b, c, d), y_{2}(a, e, j)\right)$ exists. According to Theorem 6, we first compute two cofactors with respect to $a$ and obtain $f_{\bar{a}}=1$ and $f_{a}=b c d+e j$. Note that $f_{a}$ can be easily decomposed into the form $g_{1}\left(y_{11}(b, c, d), y_{12}(e, j)\right)$, where $g_{1}\left(y_{11}, y_{12}\right)=y_{11}+y_{12}, y_{11}(b, c, d)=b c d$, and $y_{12}(e, j)=e j$. In order to obtain a nondisjoint bidecomposition of $f(X)$, we would like to represent $f_{\bar{a}}=1$ as $g_{0}\left(y_{01}, y_{02}\right)$ such that $g_{0}=g_{1}$. We choose $g_{0}\left(y_{01}, y_{02}\right)=y_{01}+y_{02}$ with $y_{01}=1$ and $y_{02}=1$. Then, combining the two decompositions, we have $f(X)=g\left(y_{1}(a, b, c, d), y_{2}(a, e, j)\right)$, where $y_{1}(a, b, c, d)=\bar{a} \cdot g_{01}+a \cdot g_{11}(b, c, d)=\bar{a}+a b c d$, $y_{2}(a, e, j)=\bar{a} \cdot g_{02}+a \cdot g_{12}(e, j)=\bar{a}+a e j$, and $g\left(y_{1}, y_{2}\right)=y_{1}+y_{2}$. To achieve efficient computation, functional decomposition operations are performed using ordered binary decision diagrams [5] in our implementation, as in approaches such as [6], [25], and [26].

## B. PLB1 in Configuration B

If PLB1 in Configuration B implements $f(X)$, a decomposition $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$ must exist for $f(X)$. There could be bridged inputs between LUTs $F$ and $G$, but $x_{H}$ does not bridge to either of them.

Let $X_{F G}=X_{F} \cup X_{G}$. We consider two cases: 1) $\left|\operatorname{cs}_{X_{F G}}(f)\right|=3,4$ or 2 ) $\left|\operatorname{cs}_{X_{F G}}(f)\right|=2$. (It is impossible that $\left|\operatorname{cs}_{X_{F G}}(f)\right|>4$ because any function $h\left(x_{H}\right)$ can only be one of zero, one, $x_{H}$, or $\bar{x}_{H}$.) In Case 1, there exists a disjoint decomposition of $f(X)$ under the bound set $X_{F G}$ with two encoding functions (according to Theorem 2). Let $f(X)=g\left(y_{1}\left(X_{F G}\right), y_{2}\left(X_{F G}\right), x_{H}\right)$ be the decomposition. Then, PLB1 in Configuration B can implement $f(X)$ if and only if the encoding functions are feasible for LUTs $F$ and $G$. This implies a partially dependent decomposition such that $\left|\sup \left(y_{1}\right)\right| \leq\left|X_{F}\right|$ and $\left|\sup \left(y_{2}\right)\right| \leq\left|X_{G}\right|$. By checking every possible input variable for $x_{H}$, we can determine the existence of a match for Case 1 based on Theorem 4.

In Case 2, a simple disjoint decomposition $g\left(y_{1}\left(X_{F G}\right), x_{H}\right)$ of $f(X)$ exists under the bound set $X_{F G}$, which does not match directly to configuration B. However, it is possible to replace a large encoding function $\left(y_{1}\left(X_{F G}\right)\right)$ with two small encoding functions. In particular, we proved that a function $f(X)=g\left(y_{1}\left(X_{F G}\right), x_{H}\right)$ can be represented as $g^{\prime}\left(z_{1}\left(X_{F}\right), z_{2}\left(X_{G}\right), x_{H}\right)$ when the following condition holds.

Theorem 7: Let $f(X)=g\left(y_{1}\left(X_{F G}\right), x_{H}\right)$ and $X_{F} \cup X_{G}=$ $X_{F G}$. A partially dependent decomposition $g^{\prime}\left(z_{1}\left(X_{F}\right)\right.$, $\left.z_{2}\left(X_{G}\right), x_{H}\right)$ of $f(X)$ exists if and only if a bidecomposition $h\left(z_{1}\left(X_{F}\right), z_{2}\left(X_{G}\right)\right)$ of $y_{1}\left(X_{F G}\right)$ exists.

Proof-Only If: Consider $f_{x_{H}}=g\left(y_{1}\left(X_{F G}\right), 1\right)$. Because $g\left(y_{1}, 1\right)=y_{1}$ or $\bar{y}_{1}$, we have $f_{x_{H}}=y_{1}\left(X_{F G}\right)$ or $\bar{y}_{1}\left(X_{F G}\right)$. On the other hand, $f_{x_{H}}=g_{x_{H}}^{\prime}\left(z_{1}\left(X_{F}\right), z_{2}\left(X_{G}\right), 1\right)=h\left(z_{1}\left(X_{F}\right)\right.$, $\left.z_{2}\left(X_{G}\right)\right)$. Therefore, $y_{1}\left(X_{F G}\right)=h\left(z_{1}\left(X_{F}\right), z_{2}\left(X_{G}\right)\right)$ or $y_{1}\left(X_{F G}\right)=\bar{h}\left(z_{1}\left(X_{F}\right), z_{2}\left(X_{G}\right)\right)$. In either case, a bidecomposition of $y_{1}\left(X_{F G}\right)$ is found.

If: Since $f(X)=g\left(y_{1}\left(X_{F G}\right), x_{H}\right)=g\left(h\left(z_{1}\left(X_{F}\right)\right.\right.$, $\left.\left.z_{2}\left(X_{G}\right)\right), x_{H}\right)=g^{\prime}\left(z_{1}\left(X_{F}\right), z_{2}\left(X_{G}\right), x_{H}\right)$, the necessary condition is proved.

Note that Cases 1 and 2 cover all possible implementations of $f(X)$ on PLB1 in Configuration B. Therefore, using the partially dependent decomposition algorithm in [13] and Theorem 7, we can determine if a wide function can be implemented on PLB1 in Configuration B. Note that $X_{F} \cap X_{G} \neq \emptyset$ is not excluded in both cases. Therefore, bridged inputs to LUTs $F$ and $G$ have been considered implicitly.

## C. PLB1 in Configuration C

If PLB1 in Configuration C implements $f(X)$, a decomposition $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$ of $f(X)$ must exist where $x_{H}$ belongs to $X_{F}$ and $X_{G}$. Besides $x_{H}$, other bridged inputs to LUTs $F$ and $G$ may exist in the implementation. We proved the following theorem for matching $f(X)$ to Configuration C.

Theorem 8: Let $x_{H} \in X_{F} \cap X_{G}, X_{1}=X_{F}-\left\{x_{H}\right\}$, and $X_{2}=X_{G}-\left\{x_{H}\right\}$. A decomposition of the form $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$ for $f(X)$ exists if and only if there exist bidecompositions $f_{\bar{x}_{H}}=g_{0}\left(y_{01}\left(X_{1}\right), y_{02}\left(X_{2}\right)\right)$ and $f_{x_{H}}=g_{1}\left(y_{11}\left(X_{1}\right), y_{12}\left(X_{2}\right)\right)$.

Proof-Only If: Define $y_{01}\left(X_{1}\right)=\left.y_{1}\left(X_{1}\right)\right|_{x_{H}=0}$, $y_{02}\left(X_{2}\right)=\left.y_{2}\left(X_{2}\right)\right|_{x_{H}=0}$, and $g_{0}\left(X_{1} \cup X_{2}\right)=g_{\bar{x}_{H}}\left(X_{1} \cup X_{2}\right)$. Then, $f_{\bar{x}_{H}}=g_{0}\left(y_{01}\left(X_{1}\right), y_{02}\left(X_{2}\right)\right)$. Similarly, define $y_{11}$ $\left(X_{1}\right)=\left.y_{1}\left(X_{1}\right)\right|_{x_{H}=1}, \quad y_{12}\left(X_{2}\right)=\left.y_{2}\left(X_{2}\right)\right|_{x_{H}=1}$, and $g_{1}\left(X_{1} \cup X_{2}\right)=g_{x_{H}}\left(X_{1} \cup X_{2}\right)$. Then, $f_{x_{H}}=$ $g_{1}\left(y_{11}\left(X_{1}\right), y_{12}\left(X_{2}\right)\right)$. The necessary condition is proved.

If: Define $y_{1}\left(X_{F}\right)=\bar{x}_{H} y_{01}\left(X_{1}\right)+x_{H} y_{02}\left(X_{2}\right)$, $y_{2}\left(X_{G}\right)=\bar{x}_{H} y_{11}\left(X_{1}\right)+x_{H} y_{12}\left(X_{2}\right)$, and $g=\bar{x}_{H} g_{0}+x_{H} g_{1}$. Then, $f(X)=\bar{x}_{H} f_{\bar{x}_{H}}+x_{H} f_{x_{H}}$ implies

$$
\begin{aligned}
f(X)= & \bar{x}_{H} g_{0}\left(y_{1}\left(X_{1}, x_{H}\right), y_{2}\left(X_{2}, x_{H}\right)\right) \\
& +x_{H} g_{1}\left(y_{1}\left(X_{1}, x_{H}\right), y_{2}\left(X_{2}, x_{H}\right)\right)
\end{aligned}
$$

A decomposition $f(X)=g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$ is then obtained.

To test the matching of a function $f(X)$ to Configurations C of PLB1, we enumerate $x_{H}$ from $X$ and compute bidecompositions for the cofactors $f_{\bar{x}_{H}}$ and $f_{x_{H}}$ such that the condition in Theorem 8 is satisfied. Because $x_{H}$ feeds to LUTs $F, G$, and $H$, the existence condition for Configuration C is less constrained compared to Configurations A and D (Section IV-D).

## D. PLB1 in Configuration D

If PLB1 in Configuration D implements $f(X)$, a decomposition $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$ of $f(X)$ must exist with $x_{H} \in$ $X_{G}$. Other bridged inputs to LUTs $F$ and $G$ may exist. We prove the following result.

Theorem 9: Let $x_{H} \in X_{G}$ and $X_{2}=X_{G}-\left\{x_{H}\right\}$. A decomposition of the form $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$ for $f(X)$ exists if and only if $f_{\bar{x}_{H}}=g_{0}\left(y_{01}\left(X_{F}\right), y_{02}\left(X_{2}\right)\right)$ and $f_{x_{H}}=$ $g_{1}\left(y_{11}\left(X_{F}\right), y_{12}\left(X_{2}\right)\right)$ such that $y_{01}\left(X_{F}\right)=y_{11}\left(X_{F}\right)$.

Proof: The proof is similar to that of Theorem 8 except that we require $y_{01}\left(X_{F}\right)=y_{11}\left(X_{F}\right)=y_{1}\left(X_{F}\right)$.

To implement a function $f(X)$ on a PLB1 in Configurations D, we select each $x_{H}$ in turn from $X$ followed by computing bidecompositions of $f_{\bar{x}_{H}}$ and $f_{x_{H}}$ to satisfy the condition in Theorem 9. Since bidecompositions of functions are not unique, for every bidecomposition $g_{0}\left(y_{01}\left(X_{F}\right), y_{02}\left(X_{2}\right)\right)$ of $f_{\bar{x}_{H}}$, we derive another bidecomposition with inverted encoding function $\bar{y}_{01}\left(X_{F}\right)$. Both bidecompositions are then tested for the condition in Theorem 9. Note that we do not derive four bidecompositions, as in the matching to Configuration A , because it is $y_{01}\left(X_{F}\right)$ and $y_{11}\left(X_{F}\right)$ under comparison rather than $g_{0}$ and $g_{1}$.

A special case that requires particular attention is when one of the cofactors is a constant (zero or one). For example, assume that $f_{\bar{x}_{H}}$ is a constant. Then, we can always obtain a decomposition of $f(X)$ by duplicating $y_{11}$ for $y_{01}$ after obtaining the bidecomposition of $f_{x_{H}}$ followed by producing $g_{0}\left(y_{01}, y_{02}\right)$ such that its value does not depend on the $y_{01}$ input (e.g., $g_{0}\left(y_{01}, y_{02}\right)=y_{02}$, where $y_{02}$ is the constant $\left.f_{\bar{x}_{H}}\right)$. Therefore, the condition in Theorem 9 is always satisfied for the special case. When both cofactors are constant, we set both $y_{01}$ and $y_{11}$ to constant and construct $g_{0}$ and $g_{1}$ accordingly to obtain a decomposition.

If there is no bridged input between LUTs $F$ and $G$ (i.e., $X_{F} \cap X_{G}=\emptyset$ ), we may also use the following theorem for efficient matching to Configuration D.

Theorem 10: Let $x_{H} \in X_{G}$ and $X_{F} \cap X_{G}=\emptyset$. A decomposition of the form $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$ for $f(X)$ exists if and only if a simple disjoint decomposition $h_{F}\left(y_{1}\left(X_{F}\right), X_{G}\right)$ and a nondisjoint decomposition $h_{G}\left(X_{F}, y_{2}\left(X_{G}\right), x_{H}\right)$ of $f(X)$ both exist.

Proof-Only If: The two decompositions $h_{F}$ and $h_{G}$ of $f(X)$ can be obtained by collapsing $y_{1}$ and $y_{2}$ into $g$, respectively. The necessary condition is proved.

If: Since $X_{F} \cap X_{G}=\emptyset$, the existence of simple disjoint decomposition $h_{F}$ of $f(X)$ implies the existence of a simple disjoint decomposition of $h_{G}$ under the bound set $X_{F}$, which in turn implies the existence of decomposition form $g\left(y_{1}\left(X_{F}\right), y_{2}\left(X_{G}\right), x_{H}\right)$ for $f(X)$. This proves the sufficient condition.

Using Theorem 10 is more efficient than using Theorem 9 for matching to Configuration D when there is no bridged inputs between LUTs $F$ and $G$. We first compute a simple disjoint decomposition under $X_{F}$ and then identify a nondisjoint variable $x_{H}$ in $X_{G}$. If both are successful, we compute the whole decomposition. Since computing a simple disjoint decomposition and identifying a nondisjoint variable can be performed efficiently, we save runtime for the case when $X_{F} \cap X_{G}=\emptyset$.

## V. Boolean Matching for PLB2

Boolean matching for PLB2 is simpler than that for PLB1 (but PLB2 is not as powerful as PLB1 for implementing wide functions as shown in Section VI). It is easy to see that bridging
$x_{H}$ to LUTs $F$ or $G$ does not help in matching wide functions to PLB2. Such a matching can be obtained by performing Shannon expansions.

Theorem 11: PLB2 can implement $f(X)$ if and only if a Shannon expansion $f(X)=\bar{x}_{H} y_{1}\left(X_{F}\right)+x_{H} y_{2}\left(X_{G}\right)$ can be obtained for some $x_{H} \in X$.

Therefore, given a wide function, we enumerate every input as the MUX selection signal and check if the supports of two cofactors contain no more than $\left|X_{F}\right|$ and $\left|X_{G}\right|$ variables, respectively. Once the constraints are met, we obtain a matching for PLB2.

## VI. Boolean Matching-Based Architecture Evaluation

We applied our Boolean matching approaches in two experiments. First, we employed them to map 1868 benchmark circuits provided by Xilinx, Inc. All circuits are known to be implementable in one XC4K CLB, but only up to $76 \%$ of them were mapped successfully with Xilinx internal tools or any other commercial FPGA tools [23]. Using Boolean matching techniques developed in this paper, we were able to achieve $100 \%$ single CLB implementation for this set of circuits. Second, we employed the matching approaches in the evaluation of four different PLB architectures (shown in Fig. 5) based on the percentage of wide functions (extracted from MCNC benchmarks) that can implemented with each PLB. Our quantitative approach can help design better FPGA logic blocks.

## A. Boolean Matching for XC4K CLB

Since $\left|X_{F}\right|=\left|X_{G}\right|=4$, wide functions are functions with six to nine variables for the XC4K CLB. We refine the XC4K CLB architecture into the configurations a to h shown in Fig. 4, with respect to the input sizes of six, seven, eight, and nine. For example, the configurations $6 . b$ and $6 . c$ are instances of Configurations A and C, respectively. Bridged inputs are explicitly shown in Fig. 4. There are as many as eight configurations for circuits with six inputs while there is only one configuration for circuits with nine inputs. Note that the configurations are not exclusive. For example, configurations 6.a, 6.b, 6.c, and 6. d are increasingly more capable in implementing six-input functions (but also require longer and longer runtime).

Refined configurations that involve multiple bridged inputs (e.g., 6.g) may combine some Configurations A, B, C, and D. Therefore, matching functions to configurations of multiple bridged inputs may require a sequence of functional decomposition operations described in Section IV. For example, configuration $6 . \mathrm{d}$ is a combination of Configurations A and C . In order to implement wide functions in configuration 6.d, we apply matching procedures for both Configuration C (based on Theorem 8) and Configuration A (based on Theorem 6). Given a function $f(X)$, denote the bridged inputs in configuration 6.d as $x_{H}$ (to all LUTs) and $x_{i}$ (to LUTs $F$ and $G$ only). According to Theorem 8, it requires that both cofactors $f_{\bar{x}_{H}}$ and $f_{x_{H}}$ have (nondisjoint) bidecompositions with $x_{i}$ being the nondisjoint variable. Consequently, we verify that both cofactors $f_{\bar{x}_{H} \bar{x}_{i}}$ and $f_{\bar{x}_{H} x_{i}}$ have (disjoint) bidecompositions that satisfy Theorem 6 and similarly for $f_{x_{H} \bar{x}_{i}}$ and $f_{x_{H} x_{i}}$. If the verification results are positive, then $f(X)$ can be


Fig. 4. Elaborated configurations of XC4K CLB. (a) Six-input configurations. (b) Seven-input configurations. (c) Eight-input configurations. (d) Nine-input configuration.
decomposed in a way that matches configuration 6.d. Clearly, matching to configuration $6 . g$ (which involves three bridged inputs) is the most time-consuming procedure, while matching to configurations without bridged inputs (6.a, 7.a, 8.a, 8.b, and 9.a) employs only simple disjoint decompositions. It is worth noting that configurations $6 . \mathrm{h}$ and 8.d are instances of Configuration B to which partially dependent decomposition is employed for a match. Since bridged inputs are taken into account implicitly in partially dependent decomposition, there is no need for computing cofactors using Shannon expansions with bridged inputs.

Initially, 20 out of 1868 Xilinx benchmark circuits had five or less inputs. After applying the SIS rugged script, we obtained an additional 292 circuits of five or less inputs. The remaining 1576 circuits consist of $393,371,423$, and 389 circuits of six, seven, eight, and nine inputs, respectively. Each circuit was then matched to the refined configurations in alphabetical order (i.e., 6. a before $6 . b$ before $6 . c$, etc.) so that implementation of the least bridged inputs could be obtained. The results are presented in Table I. Note that 30 circuits of six inputs are matched to configuration $6 . g$ which involves three bridged inputs.

TABLE I
Number of Circuits Matched to the Configurations

| Number of ciruits matched to each configuration |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 6 | 35 | 104 | 20 | 170 | 14 | 14 | 30 | 6 |
|  | 7 | 63 | 86 | 28 | 72 | 122 | 0 |  |  |
|  | 8 | 105 | 106 | 212 | 0 |  |  |  |  |
|  | 9 | 389 |  |  |  |  |  |  |  |
|  |  | .a | .b |  | $\underset{\text { nfigu }}{. d}$ | $\underset{\text { ations }}{. e}$ | .f | .g | .h |

## B. PLB Architecture Evaluation

In this section, we present our evaluation on four PLB architectures which are variations in PLB1 and PLB2 families. Their diagrams are shown in Fig. 5, where (a) XC4K CLB is $\operatorname{PLB} 1(4,4,1)$; (b) $\operatorname{PLB} 2(*, 4)$ could have different sizes of


Fig. 5. Four PLB architectures. (a) XC 4 K CLB. (b) $\operatorname{PLB} 2(*, 4)$. (c) $\operatorname{PLB} 1(1,4,1)$. (d) $\operatorname{PLB} 1(*, 4,0)$.

TABLE II
Implementation of Cuts on XC4K in Various Configurations

|  | XC4K Configurations |  |  |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
|  | A-br | A | B | C | D | all |
| 6-cuts | $51 \%$ | $90 \%$ | $98 \%$ | $91 \%$ | $49 \%$ | $99 \%$ |
| 7-cuts | $34 \%$ | $62 \%$ | $88 \%$ | $63 \%$ | $33 \%$ | $92 \%$ |

LUT $F$ but LUT $G$ has four inputs; (c) $\operatorname{PLB} 1(1,4,1)$ has a degenerated LUT $F$ of wire connection; and (d) PLB1 $(*, 4,0)$ does not have the $x_{H}$ line and LUT $G$ has four inputs.
We evaluate PLBs based on the number of wide functions (extracted from MCNC benchmarks) that each PLB can implement and the number of SRAM bits in LUTs. The bits in LUTs (for storing truth tables) are called LUT bits in the sequel. Our approach is as follows. First, we compute for each node the complete set of seven-feasible cuts [16] (where each cut corresponds to the inputs to a supernode at the node). The number of cuts largely depends on the size of the circuit. For example, there are 285,422 , and 734 instances of five cuts, six cuts, and seven cuts in $5 x p 1$, while there are 28875,65245 , and 157028 instances of five cuts, six cuts, and seven cuts in des. Second, for each cut (i.e., supernode), we compute its function and match the function to PLBs. We say a cut can be implemented by a PLB if the corresponding function can be matched to the PLB. We report the percentage of successful implementation of cuts in Tables II and III. Finally, we divide the number of implemented cuts by the number of LUT bits for each PLB to represent PLB functional capability. In other words, we measure the efficiency of LUT bits in wide function implementation. Our measurement, of course, is only one aspect of PLB architectures. Other important factors such as required routing resources are not taken into account in the evaluation. Nevertheless, we think that it is important to see the capability of various PLBs in implementing wide functions.
We match six cuts to XC4K CLB in Configurations A, B, C, and D , and report the average percentages of matched cuts in Table II. Additionally, we report the results on Configuration A while disallowing input bridging (A-br) to see its impact on wide function implementation. Comparing the results of Configurations A versus A-br as well as the results of Configura-

TABLE III
Implementation of Cuts on $\operatorname{PLB} 2(*, 4), \operatorname{PLB} 1(1,4,1)$, and PLB1 $(*, 4,0)$

|  | $\operatorname{PLB} 2\left(\left\|X_{F}\right\|, 4\right)$ |  |  | PLB1(1,4,1) |  | $\operatorname{PLB} 1\left(\left\|X_{F}\right\|, 4,0\right)$ |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $(3,4)$ | $(4,4)$ | $(5,4)$ | SD | ND | $(3,4,0)$ | $(4,4,0)$ |
| 5-cuts | $98 \%$ | $100 \%$ | $100 \%$ | $96 \%$ | $98 \%$ | $89 \%$ | $97 \%$ |
| 6-cuts | $5 \%$ | $8 \%$ | $98 \%$ | $85 \%$ | n/a | $33 \%$ | $89 \%$ |

tions $C$ versus $D$, we see the percentages of matches increase substantially when the inputs to LUTs $F, G$, and $H$ are allowed to bridge with each other. Also, we notice that each of Configurations A, B, and C alone can implement over $90 \%$ of six cuts in the MCNC benchmarks. Overall, $99 \%$ of six cuts (all) can be implemented using the XC4K CLB.

For the implementation of seven cuts on XC4K CLBs, it is interesting to see that Configuration $B$ (based on partially dependent decomposition) is more capable than other configurations (based on bidecompositions). Overall, $92 \%$ of seven cuts can be implemented using the XC4K CLB.

In the evaluation of $\operatorname{PLB} 2(*, 4)$ architectures, we consider three configurations: $\operatorname{PLB} 2(3,4), \operatorname{PLB} 2(4,4)$, and PLB2 $(5,4)$. In other words, we fixed LUT G as a 4-LUT and considered LUT F as a 3-LUT, 4-LUT, and 5-LUT, respectively. Note that it is not guaranteed that a single $\operatorname{PLB} 2(3,4)$ can realize an arbitrary five-input function. However, experimental results in Table III show that PLB2(3,4) implements 98\% of five cuts. Experiments also show that $\operatorname{PLB} 2(3,4)$ and PLB2 $(4,4)$ implement only $5 \%$ and $8 \%$ of six cuts, respectively, while PLB2 $(5,4)$ implements $98 \%$ of 6-LUTs. It is worth noting that shrinking LUT F from 4-LUT to 3-LUT loses marginally in terms of functional capability but saves substantially on LUT bits ( $25 \%$ ). Also, by expanding LUT F from 4-LUT to 5-LUT, PLB2 gains substantial capability on six-input function implementation with an additional $25 \%$ more LUT bits.

PLB1 $(1,4,1)$ has the least LUT bits among the four PLBs under evaluation. Although implementation of five cuts is not guaranteed, experiments show that $96 \%$ of five-cuts can be implemented by applying simple disjoint decomposition (SD) on the five-input functions. If nondisjoint decomposition (ND) is also applied, an additional $2 \%$ of five cuts can be implemented. PLB1 $(1,4,1)$ also implements $85 \%$ of six cuts (using SD) as well.

For $\operatorname{PLB} 1(*, 4,0)$, we considered two architectures $\operatorname{PLB} 1(3,4,0)$ and $\operatorname{PLB} 1(4,4,0)$. Although implementation of five cuts is not guaranteed, we found that $\operatorname{PLB} 1(3,4,0)$ can implement most five cuts and $1 / 3$ of six cuts, and $\operatorname{PLB} 1(4,4,0)$ can implement $97 \%$ of five cuts and $89 \%$ of six cuts.

We compared these PLBs in terms of the unit-bit implementation capability (UBIC), which is defined as the number of cuts that a PLB can implement divided by the number of LUT bits in that PLB. The comparison is presented in Table IV. Among the four evaluated PLB architectures, $\operatorname{PLB} 1(1,4,1)$ has the highest UBIC for five cuts and six cuts. In addition, we notice that $\operatorname{PLB} 2(3,4)$ has high UBIC for five cuts and $\operatorname{PLB} 1(4,4,0)$ has

TABLE IV
Functional Capability Per Each LUT Bit for PLBs

|  |  | \#cuts per LUT bit (UBIC) |  |  |
| :--- | :---: | :---: | :---: | :---: |
| PLBs | \#bits | 5-cut | 6-cut | 7-cut |
| XC4K | 40 | 1327 | 2923 | 6389 |
| PLB2(3,4) | 24 | 2167 | 249 | - |
| PLB2(4,4) | 32 | 1659 | 298 | - |
| PLB2(4,5) | 48 | 1106 | 2436 | - |
| PLB1(1,4,1) | 24 | 2167 | 4226 | - |
| PLB1(3,4,0) | 28 | 1687 | 1364 | - |
| PLB1(4,4,0) | 36 | 1430 | 2950 | - |

high UBIC for six cuts. The XC4K CLB has very high UBIC for seven cuts.

## VII. Application to Technology Mapping for FPGAs

In this section, we incorporate our Boolean matching techniques into technology mapping algorithms for depth minimization. Our mapping algorithms are targeted to the XC4K CLB ( $\operatorname{PLB} 1(4,4,1)$ ) and the XC5200 CLB (PLB2(4, 4)) architectures. However, they are applicable to general PLB1 and PLB2 types of FPGAs. We formulate the following problem.

Technology Mapping for XC4K/XC5200 Series FPGAs: Given a network $N$, compute a functionally equivalent PLB network of XC4K CLBs or XC5200 CLBs such that the depth of the PLB network is minimum.

Our approaches inherit the polynomial-time FlowMap algorithm [9] that can guarantee the minimum depth in LUT mapping solutions. Given a Boolean network $N$ of logic gates that have no more than $K$ fan-ins, FlowMap first computes the minimum level for each node $v$ in all LUT mapping solutions. This level is called the LUT label of $v$, denoted as lut_label $(v)$ in this paper. After computing the labels, FlowMap generates a mapping solution based on them.

Computing node labels is the key operation in FlowMap and is briefly reviewed as follows. Every PI has a minimum level of zero. Remaining node labels are computed from PIs to POs in a topological order. Let $N_{v}$ denote the subnetwork rooted at node $v$. A cut in $N_{v}$ is a set of nodes that separates $v$ from PIs. A cut is $K$ feasible if the node cutset contains at most $K$ nodes. The height of a cut is defined as the largest label for nodes in the cut. Let $p$ be the largest label among the fan-ins of $v$. It was shown in [9] that lut_label $(v)=p$ if there is a $K$-feasible cut of height $p-1$ in $N_{v}$, which can be verified using the max-flow min-cut algorithm. If such a cut cannot be found, then lut_label $(v)=p+1$. Our mapping algorithms take similar steps to compute the minimum depth in CLB networks.

## A. Technology Mapping for XC4K FPGAs

In parallel to the definition of LUT label, the minimum level of node $v$ in any CLB network is called the CLB label of $v$, denoted clb_label $(v)$. In general, clb_label $(v) \leq$ lut_label $(v)$.

The largest CLB label in a network $N$ is called the CLB depth CLB_depth $(N)$ of the network.

Our mapping algorithm for the XC 4 K CLB is called BM-Map. Before mapping, the input network is decomposed into a two-bounded network. BM-Map has three phases: initialization, labeling, and mapping. In initialization, BM-Map computes the set $C_{5}(v)$ of all five-feasible cuts for every node $v \in N$. (Wide cuts are generated using five-feasible cuts for runtime consideration). In the labeling phase, BM-Map computes both lut_label $(v)$ and clb_label $(v)$ using the set $C_{5}(v)$. Nodes are proceeded in a topological order in both procedures. Finally, BM-Map produces a XC4K CLB network in the mapping phase.

The cutset $C_{5}(v)$ is obtained using cut enumeration techniques in [16]. It computes $K$-feasible cuts in $N_{v}$ by merging the cuts of the fan-ins of $v$ and rejecting those cuts that are not $K$ feasible. In theory, the number of $K$-feasible cuts grows exponentially with respect to $K$. However, for $K \leq 5$, this computation is efficient in practice. For most benchmarks, we see about 30-70 five-feasible cuts per node.

Let $p$ be the largest LUT label among the fan-ins of $v$. Instead of using max-flow min-cut procedures, BM-Map determines if lut label $(v)=p$ by looking for a cut of height $p-1$ in the cutset $C_{\tilde{5}}(v)$. To compute the clb_label $(v)$, let $q$ be the largest CLB label among the fan-ins of $v$. BM-Map performs the following two checks.
(C1) If there exists a $K$-feasible cut of height $q-1$ in $C_{5}(v)$, then clb_label $(v)=q$.
(C2) Otherwise, if there exists a nine-feasible cut of height $q-1$ of which the corresponding wide function can be successfully matched to the XC 4 K CLB, then $\operatorname{clb}$ label $(v)=q$.
If both checks fail, then $\operatorname{clb} \operatorname{label}(v)=q+1$.
We take two actions to save runtime in label computation. First, we do not exhaust all XC4K configurations in Boolean matching. We only test configurations (in Fig. 4) 6.a and 6.c for six cuts, 7.c and 7.f for seven cuts, 8.c for eight cuts, and 9.a for nine cuts because they have high matching percentages (Table II) with relatively low decomposition complexity. Second, to obtain wide cuts (of six to nine nodes) at each node $v$, we simply merge the cuts in $C_{5}\left(u_{1}\right)$ and $C_{5}\left(u_{2}\right)$, where $u_{1}$ and $u_{2}$ are fan-ins of $v$. By doing so, we trade the completeness of nine-feasible cuts for runtime.

After every node in the network has been labeled, BM-Map generates a CLB mapping solution that respects labels. A PO node $v$ is a critical PO node if LUT_label $(v) \geq$ CLB_depth $(N)$. For those critical POs and their fan-in networks, BM-Map covers them with CLBs to guarantee the CLB depth computed in the labeling phase. For the remaining noncritical POs and their fan-in networks, BM-Map covers them with LUTs to save area. At last, BM-Map packs LUTs into CLBs using an efficient procedure match_4k proposed in [15]. The BM-Map algorithm is outlined in Fig. 6.

## B. Technology Mapping for XC5200 FPGAs

While the XC 4 K CLB can implement a large number of six cuts and seven cuts, it was shown in previous sections that the

```
function BM-Map ( N)
/* N = input network */
Compute C}\mp@subsup{C}{5}{}(v)\mathrm{ for every v}v\in
Compute lut label(v) for every v\inN
Compute clb label (v) for every v\inNusing (C1) and (C2)
    in Section 7.1 (matched wide cuts saved)
CLB depth (N)=max{CLB label (v)|v\inN}
for each critical PO v}\mathrm{ with LUT label (v) }\geqCLB depth(N) d
    Get the cut of height clb label(v)-1
    Mapv to a CLB with inputs from the cut
    Map every }u\mathrm{ in the cut recursively using CLBs
for each remaining non-critical PO v do
    Select a cut from C55(v)
    Map v to an LUT with inputs from the cut
    Map every }u\mathrm{ in the cut recursively using LUTs
Pack LUTs into CLBs using match 4k
```

Fig. 6. BM-Map algorithm.

XC5200 CLB implements only a very small percentage of them. However, we can exploit the XC5200 CLB architecture based on an interesting result observed in our experiments.

Let us first introduce two concepts. A wide function $f(X)$ is fully implementable on a XC5200 CLB if each cofactor $f_{x_{i}}$ and $f_{\bar{x}_{i}}$ depends on at most four variables for some $x_{i} \in X$ (Theorem 11), while $f(X)$ is partially implementable if only one of the two cofactors satisfies the condition. In Table V, the columns titled "fit" and " $1 / 2$-fit" show the number of cuts that are fully and partially implementable on XC5200 CLBs, respectively. Although very few six cuts and seven cuts are fully implementable, most of them are partially implementable. Based on this observation, we consider using XC5200 CLBs for partially implementable functions. [Note that most wide functions are fully implementable on XC4K CLBs (see Table II). As a result, there is little benefit to consider partially implementable functions.]

Our mapping algorithm for XC5200 CLBs, called BMD-Map, enhances BM-Map by incorporating functional decomposition $(D)$ operations into label computation. In initialization, BMD-Map computes the cutset $C_{7}(v)$ of all seven-feasible cuts for every node $v \in N$. (We can afford testing more wide cuts since matching to XC5200 CLB is much less time consuming.) In the labeling phase, BMD-Map determines if clb_label $(v)=q$ using the following three checks.
(C1) If there exists a five-feasible cut of height $q-1$ in $C_{7}(v)$, then clb_label $(v)=q$.
(C2) Otherwise, if there exists a wide cut of height $q-1$ in $C_{7}(v)$ of which the corresponding wide function is full implementable, then clb_label $(v)=q$.
(C3) Otherwise, if there exists a wide cut of height $q-1$ in $C_{7}(v)$ of which the corresponding wide function is partially implementable and the infeasible cofactor can be decomposed to obtain a new cut of height $q-1$ (ex-

TABLE V
Matching Wide Functions to XC5200 CLBs

| Circuits | 6 -cuts | fit | $1 / 2$-fit | 7 -cuts | fit | $1 / 2$-fit |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| 5 xp1 | 422 | 186 | 233 | 734 | 159 | 563 |
| 9 sym | 1256 | 235 | 1020 | 2333 | 143 | 2173 |
| 9 symml | 1156 | 177 | 978 | 2195 | 84 | 2080 |
| C499 | 9716 | 370 | 8826 | 27599 | 636 | 22601 |
| C880 | 3969 | 210 | 3727 | 9079 | 75 | 8858 |
| alu2 | 6666 | 573 | 6026 | 18231 | 482 | 17494 |
| alu4 | 14841 | 1140 | 13376 | 40332 | 604 | 38845 |
| apex6 | 2691 | 253 | 2437 | 4370 | 152 | 4216 |
| apex7 | 743 | 76 | 657 | 1342 | 39 | 1283 |
| count | 94 | 15 | 79 | 87 | 0 | 87 |
| des | 65245 | 5177 | 59192 | 157028 | 2903 | 147891 |
| duke2 | 4606 | 180 | 4426 | 10106 | 151 | 9955 |
| misex1 | 266 | 57 | 209 | 400 | 19 | 381 |
| rd84 | 2351 | 209 | 2142 | 5307 | 50 | 5198 |
| rot | 4857 | 335 | 4410 | 10362 | 177 | 9980 |
| vg2 | 410 | 32 | 378 | 850 | 29 | 806 |
| z4ml | 36 | 0 | 35 | 36 | 0 | 34 |
| total | 119325 | 9225 | 108151 | 290391 | 5703 | 272445 |
| ratio | 1 | $8 \%$ | $91 \%$ | 1 | $2 \%$ | $94 \%$ |

plained in the next paragraph), then $\operatorname{clb} \operatorname{label}(v)=q$. We refer to the decomposition of the infeasible cofactor as a decomposition operation.
Otherwise, $\operatorname{clb}$ label $(v)=q+1$.
We illustrate the decomposition operation (C3) in Fig. 7. Let $X=\left\{x_{1}, \ldots, x_{6}\right\}$ be a six cut of $v$ and all nodes in $X$ have a label $q-1$. Let $f(X)$ denote the function of the subnetwork rooted at $v$ with inputs from $X$ [see Fig. 7(a)]. Let $f(X)=$ $\bar{x}_{1} F\left(X_{F}\right)+x_{1} I\left(X_{I}\right)$ and assume that $\left|X_{I}\right|=5$. Then, $f(X)$ is partially implementable on XC5200 CLB [see Fig. 7(b)]. The dash line represents possible bridged inputs to LUTs. Decomposition operations are performed iteratively in three steps.

1) Choose a two-variable bound set $B \subset X_{I}$.
2) Perform a simple disjoint decomposition $I\left(X_{I}\right)=$ $G\left(x_{7}(B), X_{I}-B\right)$ under the bound set $B$.
3) If the decomposition is successful, create a node $u$ for $x_{7}(B)$ [see Fig. 7(c)] and compute a min-cut of height $q-2$ in the cone $N_{u}$ rooted at $u$. If the min-cut is $K$ feasible, then $l(u)=q-1$ and, therefore, $l(v)=q$ since all the inputs of $F, G$, and $H$ have a label $q-1$.
Steps 1 to 3 are iterated for all possible bound sets until a success is found or bound sets are exhausted. In the latter case, we assign $l(v)=q+1$.

## VIII. Technology Mapping Experimental Results

We conducted experiments on a Sun ULTRA2 workstation with 256 MB of memory. The tested circuits are MCNC benchmarks which were optimized using the SIS rugged script [32] and decomposed into two-input networks using the dmig algorithm [7], [35]. The mapping goal was to minimize the CLB depth with consideration to area minimization.


Fig. 7. Decomposition operation in step (S3) for XC5200 CLB. (a) Six-input supernode. (b) Partial implementation on XC5200 CLB. (c) Implementation after decomposition operation.

TABLE VI
BM-Map Mapping Results for XC4K CLBs

|  |  |  |  |  | FlowMap |  |  | BM-Map(10) |  |  | BM-Map(50) |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Circuits | D | CLB | T(s) | D | CLB | T(s) | D | CLB | T(s) |  |  |  |  |
| 5xp1 | 3 | 17 | 0.5 | 2 | 19 | 4.1 | 2 | 19 | 5.1 |  |  |  |  |
| 9sym | 5 | 35 | 1.1 | 4 | 39 | 11.9 | 4 | 39 | 20.0 |  |  |  |  |
| 9symml | 5 | 35 | 1.1 | 4 | 38 | 15.2 | 4 | 39 | 23.6 |  |  |  |  |
| C499 | 4 | 34 | 4.7 | 4 | 63 | 21.6 | 4 | 63 | 34.2 |  |  |  |  |
| C880 | 8 | 65 | 3.0 | 7 | 89 | 15.8 | 7 | 91 | 89.7 |  |  |  |  |
| alu2 | 9 | 90 | 3.8 | 7 | 108 | 53.9 | 6 | 96 | 97.8 |  |  |  |  |
| alu4 | 9 | 144 | 7.4 | 8 | 204 | 95.2 | 8 | 199 | 181.6 |  |  |  |  |
| apex6 | 5 | 155 | 4.9 | 4 | 184 | 31.3 | 4 | 186 | 34.6 |  |  |  |  |
| apex7 | 4 | 40 | 1.4 | 3 | 46 | 18.4 | 3 | 46 | 15.4 |  |  |  |  |
| count | 5 | 19 | 0.7 | 4 | 24 | 2.7 | 4 | 24 | 2.6 |  |  |  |  |
| des | 5 | 672 | 32.9 | 5 | 978 | 746.4 | 4 | 764 | 1180.8 |  |  |  |  |
| duke2 | 4 | 93 | 2.8 | 4 | 118 | 21.7 | 4 | 118 | 47.3 |  |  |  |  |
| misex1 | 2 | 10 | 0.3 | 2 | 11 | 1.6 | 2 | 11 | 2.1 |  |  |  |  |
| rd84 | 4 | 34 | 1.2 | 3 | 44 | 7.3 | 3 | 45 | 14.7 |  |  |  |  |
| rot | 7 | 130 | 5.0 | 6 | 156 | 29.5 | 6 | 159 | 44.4 |  |  |  |  |
| vg2 | 3 | 15 | 0.4 | 3 | 20 | 14.2 | 2 | 23 | 20.1 |  |  |  |  |
| z4ml | 2 | 3 | 0.2 | 2 | 3 | 0.5 | 2 | 3 | 0.6 |  |  |  |  |
| total | 84 | 1591 | 71.4 | 72 | 2144 | 1091.3 | 69 | 1925 | 1814.6 |  |  |  |  |
| arith.mean | 1.00 | 1.00 | 1.00 | 0.86 | 1.35 | 15.28 | 0.82 | 1.21 | 25.41 |  |  |  |  |
| geo.mean | 1.00 | 1.00 | 1.00 | 0.86 | 1.25 | 8.41 | 0.82 | 1.24 | 13.18 |  |  |  |  |

In the first experiment, we applied FlowMap and BM-Map to map MCNC benchmarks into XC4K CLBs. After technology mapping, match_4k [15] was employed to pack LUTs into XC4K CLBs. We limited the maximal number of wide cuts to ten and 50 in BM-Map(10) and BM-Map(50) and reported corresponding results in Table VI. Compared to FlowMap, BM-Map obtained $14 \%$ and $18 \%$ smaller depth when ten and 50 wide cuts were tested, respectively. However, BM-Map uses substantially more CLBs compared to FlowMap. After careful examination, we found that a large percentage of 5-LUTs mapped by FlowMap were decomposed by match_4k into 2-LUTs and 4-LUTs and subsequently packed into CLBs with other 4-LUTs. This is very efficient for area minimization. This benefit does not happen to BM-Map because it uses LUTs only to cover noncritical portions of the input network and obtains much less 5-LUTs. In general, the more cuts that are tested
for matching in BM-Map, the better the mapping results will be, but the longer the runtime. The ratio, however, is biased significantly by the circuit des.

In the second experiment, we applied our mapping approach to the technology mapping for XC5200 FPGAs. After mapping, LUTs were packed into XC5200 CLBs. One CLB is allocated for one 5-LUT as well as for one pair of 4-LUTs. We compared BMD-Map with the CutMap algorithm [12]. CutMap also inherits the FlowMap algorithm, but in addition performs simultaneous area minimization. It obtains the same LUT depth as obtained by FlowMap, but uses $18 \%$ less 5-LUTs on average for industrial benchmarks [22]. The mapping results are reported in Table VII. Compared to CutMap, BM-Map obtains $7 \%$ smaller depth with $15 \%$ smaller area, and BMD-Map obtains $12 \%$ smaller depth with $6 \%$ larger area. The runtime of BM-Map and BMD-Map are both one order of magnitude

TABLE VII
BMD-MAP Mapping Results for XC5200 CLBs

|  | CutMap |  |  | BM-Map |  |  | BMD-Map |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Circuits | D | CLB | T(s) | D | CLB | T(s) | D | CLB | T(s) |
| 5xp1 | 3 | 18 | 0 | 3 | 19 | 3 | 3 | 19 | 2 |
| 9sym | 5 | 48 | 1 | 4 | 43 | 10 | 4 | 46 | 9 |
| 9symml | 5 | 46 | 1 | 4 | 45 | 10 | 4 | 48 | 8 |
| C499 | 4 | 53 | 7 | 4 | 50 | 323 | 4 | 50 | 272 |
| C880 | 8 | 101 | 3 | 8 | 89 | 45 | 8 | 89 | 43 |
| alu2 | 9 | 118 | 4 | 8 | 106 | 148 | 7 | 124 | 132 |
| alu4 | 9 | 194 | 11 | 9 | 184 | 374 | 8 | 206 | 327 |
| apex6 | 5 | 186 | 7 | 4 | 165 | 22 | 4 | 175 | 19 |
| apex7 | 4 | 56 | 1 | 3 | 52 | 6 | 3 | 57 | 6 |
| count | 5 | 28 | 0 | 5 | 23 | 1 | 4 | 28 | 1 |
| des | 5 | 925 | 116 | 5 | 714 | 1750 | 5 | 1040 | 1588 |
| duke2 | 4 | 131 | 4 | 4 | 121 | 70 | 4 | 129 | 66 |
| misex1 | 2 | 12 | 0 | 2 | 9 | 2 | 2 | 9 | 1 |
| rd84 | 4 | 40 | 2 | 4 | 37 | 45 | 4 | 40 | 37 |
| rot | 7 | 166 | 6 | 6 | 154 | 62 | 5 | 188 | 56 |
| vg2 | 3 | 19 | 0 | 3 | 18 | 4 | 3 | 18 | 5 |
| z4ml | 2 | 4 | 0 | 2 | 4 | 0 | 2 | 4 | 0 |
| total | 84 | 2145 | 163 | 78 | 1833 | 2875 | 74 | 2270 | 2572 |
| arith.mean | 1.00 | 1.00 | 1.00 | 0.93 | 0.85 | 17.64 | 0.88 | 1.06 | 15.78 |
| geo.mean | 1.00 | 1.00 |  | 0.93 | 0.91 |  | 0.90 | 0.99 |  |

longer than CutMap. Comparing BM-Map with BMD-Map, we can see the impact of functional decomposition in XC5200 mapping.

## IX. CONCLUSION

We have presented new Boolean matching methods for LUT-based PLBs and their applications to architecture evaluation and FPGA technology mapping. Our Boolean matching methods employ functional decomposition operations to represent functions in forms corresponding to the target PLB architecture. Existence conditions for new functional decomposition forms are given and proved. We applied the methods to the evaluation of PLB architectures in terms of logic implementation capability. Experimental results show that the Xilinx XC4K CLB can implement $98 \%$ and $88 \%$ of six- and seven-variable functions extracted from MCNC benchmarks, respectively, while a simplified PLB architecture implements the largest amount of functions per LUT bit. We developed new technology mapping algorithms that employ the Boolean matching techniques for depth minimization. Compared to conventional LUT mapping approaches, experimental results show $18 \%$ and $12 \%$ depth reduction on average for the Xilinx XC4K series and XC5200 series FPGAs, respectively, with up to $15 \%$ area reduction in XC5200 FPGAs. Our Boolean matching techniques can be useful for designing future FPGA architectures and better utilization of FPGA silicon resources.

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Jason (Jingsheng) Cong (S'88-M'90-SM'96F'01) received the B.S. degree in computer science from the Peking University, Beijing, China, in 1985, and received the M.S. and Ph.D. degrees in computer science from the University of Illinois, Urbana-Champaign, in 1987 and 1990, respectively.

He is currently a Professor in the Computer Science Department and Co-Director of the VLSI CAD Laboratory at University of California, Los Angeles. His research interests include computer-aided design of VLSI circuits and systems, rapid prototyping of computer and communication systems, design and analysis of algorithms, and computer architecture.
Dr. Cong received the Best Graduate Award from Peking University in 1985, the Ross Martin Award for Excellence in Research from the University of Illinois in 1989, the National Science Foundation Research Initiation Award and Young Investigator Award in 1991 and 1993, respectively, the Northrop Corporation Outstanding Junior Faculty Research Award from University of California, Los Angeles, in 1993, the ACM SIGDA Meritorious Service Award in 1998, and the IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems Best Paper Award in 1995. He is an Associate Editor of ACM Transactions on Design Automation of Electronic Systems.


Yean-Yow Hwang received the B.S. degree in electrical engineering from the National Taiwan University, Taipei, Taiwan, in and the M.S. and Ph.D. degrees in computer science from the University of California, Riverside and Los Angeles, in 1987 and 1999, respectively.
He was a Research Engineer at Synopsys, Inc. and is currently a Technical Staff Engineer at the Altera Corporation, San Jose, CA. His research interests include algorithms, logic synthesis, postlayout optimization, VLSI design tools, and FPGA
architectures.


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    J. Cong is with the Computer Science Department, University of California, Los Angeles, CA 90024 USA.
    Y.-Y. Hwang is with the Altera Corporation, San Jose, CA 95134 USA.

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