# BDS: A BDD-Based Logic Optimization System 

Congguang Yang Maciej Ciesielski<br>Dept. of Electrical \& Computer Engineering University of Massachusetts, Amherst, MA<br>cyang,ciesiel@ecs.umass.edu

Vigyan Singhal<br>Tempus Fugit, Inc. Albany, CA 94706<br>vigyan@home.com


#### Abstract

This paper describes a new BDD-based logic optimization system, $B D S$. It is based on a recently developed theory for BDD-based logic decomposition, which supports both algebraic and Boolean factorization. New techniques, which are crucial to the manipulation of BDDs in a partitioned Boolean network environment, are described in detail. The experimental results show that $B D S$ has a capability to handle very large circuits. It offers a superior runtime advantage over SIS, with comparable results in terms of circuit area and often improved delay.


## 1. INTRODUCTION

Through the continuously intensive research and development in logic synthesis area for the last twenty years, the general framework for logic synthesis has been well established. While the space for further improvement of the synthesis flow seems to be limited, there is still potential for significant improvement in many procedures in the synthesis process [1]. This is especially true when more efficient ways to represent Boolean functions become available.

The history of logic synthesis demonstrates a simple, yet clear fact that the representation of Boolean logic plays a central role in the evolution of synthesis methods. It seems natural that logic synthesis methods will continue evolving with the advent of newer and more efficient Boolean logic representations. We believe that this evolution will be further enabled by the accumulation of expertise in Binary Decision Diagrams (BDDs). Our research is trying to address this new opportunity.

BDDs were first proposed by Akers [2], and popularized by Bryant [3] and Brace et al [4]. Due to their implicit power to represent Boolean functions, BDD's are considered the most efficient Boolean representation known so far. The synthesis approaches based on decomposition and manipulation of BDD's have been found promising in FPGA synthesis [5, $6,7]$, PTL synthesis $[8,9,10]$, multi-level synthesis [11, 12, $13,14,15]$, and mixed CMOS/PTL synthesis [16, 17]. All these BDD-based synthesis methods have a great potential

Permission to make digital/hardcopy of all or part of this work for personal or classroom use is granted without fee provided that copies are not made or distributed for profit or commercial advantage, the copyright notice, the title of the publication and its date appear, and notice is given that copying is by permission of ACM, Inc. To copy otherwise, to republish, to post on servers or to redistribute to lists, requires prior specific permission and/or a fee.
DAC 2000, Los Angeles, California
(c) 2000 ACM 1-58113-188-7/00/0006.. $\$ 5.00$
to outperform traditional logic synthesis approaches.
In [18] we proposed a new BDD-based logic decomposition method based on dominators. Our work was based on the observation that the structure of a properly ordered BDD reveals important information about its functional decomposition. Different structures, called dominators can be identified in a BDD, leading to efficient AND, OR, XOR and MUX decompositions. This method offers a capability to efficiently perform both algebraic and Boolean factorizations. In terms of the area and runtime, the synthesis results obtained with this method for AND/OR-intensive functions are comparable to that of SIS, while results for XOR-intensive functions are significantly better than SIS. In both cases, our program demonstrates great runtime advantage over SIS. However, the method is limited to the decomposition of monolithic (global) BDDs, and as such it can only be used to optimize circuits for which BDDs can be constructed.

In this paper, a complete BDD-based logic optimization system, $B D S$, designed to handle arbitrarily large circuits, is described. While retaining the BDD decomposition capability of [18], a more general framework which incorporates a typical logic synthesis procedures has been implemented. Several techniques which proved very efficient in manipulating BDDs in the partitioned Boolean network environment are discussed in detail.

## 2. SYNTHESIS FLOW

A very important feature of a logic synthesis system is its scalability. The scalability requires that the size of the representation of a problem be proportional to the size of the problem itself. In our case, the size of a BDD should be proportional to the size of a circuit (which is commonly measured by the number of gates). However, the size of a global BDD for a given Boolean network is completely unpredictable. It strongly depends on the type of the circuit, and not necessarily on the total number of gates. Therefore, a proper partitioning of Boolean network is required prior to performing the BDD decomposition. Fortunately, the same problem appeared in traditional logic synthesis, where it has been handled properly.

Current multi-level logic synthesis flow exemplified by SIS has drawn from twenty years of intensive research. We believe it has the capability to handle very large circuits and it does grasp the essence of logic synthesis in general. Therefore, $B D S$ adopts the general synthesis flow of SIS. Fig. 1 compares the synthesis flow of SIS and BDS. The similarity between the two systems is obvious. The fundamental difference between SIS and BDS is the way
in which they represent Boolean nodes and carry out all individual synthesis procedures. In BDS, after a Boolean network has been built all Boolean nodes are represented as local BDDs (using only immediate fanins for each node). All the subsequent procedures are carried out based on the local BDDs.

It should be mentioned that all procedures in the synthesis flow are heavily influenced by the underlying Boolean representation. Therefore, while retaining a similar synthesis flow, new algorithms specially tailored for BDDs should be developed for all the procedures. In the following section, the BDD decomposition techniques introduced in [18] are briefly reviewed. These techniques are central to $B D S$. Other essential procedures in the synthesis flow are presented next.


Figure 1: Synthesis flow of SIS and BDS

## 3. REVIEW OF BDD DECOMPOSITION

The BDD-based logic decomposition techniques used by BDS are summarized in Table 1. Detailed explanation of the terminology and examples can be found in [18]. All types of atomic decompositions and their corresponding BDD structures (dominators) have been identified. These dominators form the central BDD decomposition engine of $B D S$. The decomposition begins with the BDD structure scan. The BDD structural information obtained by the scan is used to guide the various decomposition types.

| Type | BDD Structure | Decomposition |
| :---: | :--- | :--- |
| 1 | 1-dominator | algebraic AND |
| 2 | 0-dominator | algebraic OR |
| 3 | x-dominator | algebraic XNOR |
| 4 | generalized dominator | Boolean AND/OR |
| 5 | generalized $x$-dominator | Boolean XNOR |
| 6 | cof. w.r.t. single node | simple MUX |
| 7 | cof. w.r.t. super node | complex MUX |

Table 1: Various dominators used in BDD decomposition

The BDD decomposition process is recorded by constructing a set of factoring trees for each function. A factoring tree will keep growing until the BDD decomposition is completed. Subsequently, several steps can be applied to the factoring
trees to further improve the decomposition results. In particular, sharing between different factoring trees can be efficiently detected. For this purpose, BDDs are constructed for all factoring trees in a bottom-up fashion, and canonicity property of a BDD is used to identify functionally equivalent sub-trees.

### 3.1 BDD Decomposition - Review

The following example, for function $F=e+b d$, shown in Fig. 2(a), will illustrate the idea of Boolean AND/OR decomposition based on a general dominator. First, a cut is performed on the BDD of $F$. The generalized dominator is the portion of the BDD above the cut. The edges of the generalized dominator that that are connected to terminal nodes in the original BDD remain connected to those terminals, while the remaining edges (in this case a 1edge incident to node $d$ ) are redirected to constant 1 . This results in a Boolean divisor $D=e+d$. The quotient $Q$ is obtained from $F$ by minimizing $F$ using the off-set $\left\{e^{\prime} d^{\prime}\right\}$ of $D$ as a don't care. This leads to $Q=e+b$. The final decomposition is $F=(e+d)(e+b)$. It can be shown that all cuts form a set of compatible classes, each class leading to a unique decomposition (division). This feature drastically reduces the number of cuts to be considered.


Figure 2: A simple example of Boolean division.
The following example illustrates additional BDD decomposition capabilities of our tool, namely the XOR and MUX decompositions. Given an initial global BDD representation of a two-output function, $\{g, h\}$, the BDD for each primary output is decomposed independently of each other, and the factoring trees constructed for each function. Finally, logic sharing is identified in the resulting factoring trees. Fig. 3 shows the decomposition of BDD $g$. First, an $x$-dominator of $g$ is found, leading to an algebraic XNOR decomposition. $X$ dominator is a BDD structure which contains a node (in our example node $d$ ) with the property that all paths from root to terminal nodes $\mathbf{1}$ and $\mathbf{0}$ pass through that node. In this case $F$ can be decomposed into $F=u \bar{\oplus} f$, where $f$ is a function rooted at $d$, and $u$ is obtained from the original BDD by replacing $f$ with 1 . A factoring tree node, with operator XNOR, is created to record this decomposition.

Fig. 4 shows the decomposition of function $h$ based on the extraction of complex MUX. Such a MUX decomposition exists if the BDD contains two nodes that cover all paths from


Figure 3: Decomposition of BDD $g$.
root to terminal nodes. In our case these are the $d$ nodes below the cut. The function can then be decomposed into $F=v^{\prime}(d e)+v(d+e)$, where $v$ is derived from the portion of the BDD above the cut, by redirecting the two $d$ nodes to 0 and 1, respectively. Additional MUX decomposition is possible for function $v$. A factoring tree node, with operator MUX, is created to record this decomposition.


Figure 4: Decomposition of BDD $h$.
Finally, logic sharing between functions $g$ and $h$ is efficiently identified by constructing BDDs in a bottom-up fashion on the factoring trees for both functions. Fig. 5 shows the process of extracting the shared logic. The shadowed subtrees of $g$ and $h$ are functionally equivalent. This equivalence can be identified by constructing BDDs on both factoring trees.


Figure 5: Sharing extraction on the factoring trees of $g$ and $h$.

### 3.2 BDD Store/Load Mechanism

In our BDD-based logic optimization scheme [18], BDD variable reordering algorithm serves as an implicit logic simplification. It should be emphasized that variables are reordered with respect to a BDD manager, not w.r.t. a specific BDD. Hence, if there is more than one BDD in the manager, variable reordering may not result in the desired simplification. In order to achieve maximum logic simplification of a Boolean function (BDD), all other BDDs must be freed from this BDD manager before performing variable reordering. However, those "freed" BDDs must be present in the BDD manager when they are needed for decomposition at a later time. Therefore, an efficient store/load mechanism must be developed.

A naive way to store BDDs is to represent them in a SOP form. The advantage of SOP form is that the BDD can be reconstructed under a variable order which is different from the order when the BDD is stored. This offers some flexibility for the implementation. However, since the number of SOP terms of a BDD could be exponential in the number of BDD nodes, storing BDDs in SOP form is not a feasible solution.

A new data structure, bddPool, has been devised to perform BDD store/load operations. Basically, a bddPool is a DAG which is graphically isomorphic to the BDD it represents. A BDD is copied to a bddPool before it is freed from the BDD manager. The BDD can be reconstructed later by applying an ite operation $|V|$ times, where $|V|$ is the number of BDD nodes. Since an ite operation takes constant time, the overall complexity of our BDD store/load algorithm is $O(|V|)$. The disadvantage of bddPool is that the variable order of the BDD manager into which a BDD is loaded must be the same as the order in which a BDD is stored. Forcing a BDD manager to a certain variable order could result in an exponential increase in the BDD size if the manager is not empty. However, in our application, when a BDD is loaded, the BDD manager is always empty. A complete process of store/load is shown in Fig 6.

Another important feature of our bddPool mechanism is to allow the variable substitution during the process of BDD reconstruction. This can be accomplished easily by modifying the ite operator as $f=\operatorname{ite}(\mathcal{M}(x), g, h)$, while $\mathcal{M}$ is the mapping of variables. This feature plays a crucial role in our efficient iterative eliminate paradigm (Section 4.2).


Figure 6: BDD store/load mechanism. (a) A BDD. (b) BDD stored in bddPool. (c) Re-construction of BDD from bddPool in a bottom-up fashion.

## 4. IMPLEMENTATION ISSUES

### 4.1 Sweeping Boolean Networks

Procedure sweep is the first step in the proposed synthesis flow. It removes some obvious redundancy from the Boolean networks. There is no real logic optimization involved in this procedure. However, for some multi-level Boolean functions, sweep plays a very important role in removing redundancy from the networks.

### 4.1.1 Constant and Single-Variable Nodes Removal

A logic optimization program should take advantage of constant and single-variable nodes to reduce the complexity of a Boolean network. It should be noted that the removal
of one constant or single-variable node may create another, and such nodes may be produced during the process of logic optimization. Therefore, sweep is iterative, and it is invoked many times in a complete logic optimization process.

### 4.1.2 Removal of Functionally Equivalent Nodes

Traditionally, when a multi-level Boolean function is represented as a SOP form, only constant and single-variable nodes can be identified and removed from the Boolean network during sweep. When a Boolean network is represented in the BDD form, there is an added bonus. Since BDD is canonical, the functional equivalence between different Boolean nodes can be detected easily. Therefore, in addition to removing constant and single-variable nodes, all functionally equivalent nodes can be also be removed from a Boolean network during sweep.

Although the functionally equivalent Boolean nodes in a Boolean network can also be removed by later optimization procedures (e.g. eliminate, re-substitution) in the traditional synthesis flow, it is always beneficial to remove this type of redundancy before the actual logic optimization. Shown in Table 2 is the number of functionally duplicated nodes for some famous testcases. We were surprised to find so much redundancy in those testcases. This is the first time ever that functionally duplicated Boolean nodes can be removed before actual logic optimization.

| Circuit | Total Nodes | Duplicated Nodes |
| :--- | :---: | :---: |
| C1908 | 441 | 118 |
| C2670 | 787 | 72 |
| C3540 | 956 | 247 |
| C5315 | 1467 | 197 |
| C6228 | 2353 | 30 |
| C7552 | 2165 | 355 |
| C880 | 302 | 10 |
| dalu | 985 | 249 |
| i8 | 1183 | 186 |
| i9 | 329 | 22 |
| i10 | 1634 | 84 |
| pair | 830 | 16 |
| vda | 123 | 3 |

Table 2: Number of functionally duplicated nodes in a Boolean network

Since the removal of one functionally equivalent node may create another, the duplication removal in $B D S$ is iterative. The numbers shown in Table 2 are just the number of duplicated nodes in the first iteration. Removing functionally duplicated Boolean nodes helps BDS to reach the final optimized netlist. This also contributes to the runtime advantage over traditional approaches, because logic optimization algorithms are generally more expensive than sweep.

### 4.2 Node Elimination

To comply with the scalability requirement, a Boolean network should be allowed to be partially collapsed into a set of super Boolean nodes; then logic optimization algorithms can be applied on each super node. Partial collapsing also helps to remove logic redundancy which is embedded in the multi-level configuration. The most frequent cause of
redundancy is the so-called re-convergence. Partial collapsing is very efficient at removing it.

To carry out partial collapsing, procedure such as eliminate comes into play. eliminate attempts to find a partially collapsed Boolean network such that Boolean nodes are not too large for logic optimization algorithms. On the other hand, Boolean nodes should not be too fine, otherwise some redundancy may remain in the network. A properly designed eliminate scheme provides a better starting point for logic optimization algorithms.
eliminate has been successfully implemented in SIS [19]. Cost estimation based on the factored forms was used to guide the process of elimination. Two approaches have been proposed for eliminate using BDDs. The first one is based on progressive elimination [9]. In this approach, BDDs are constructed from primary inputs to primary outputs. At any point, if the size of a BDD is larger than a predefined threshold, an intermediate variable is introduced and the BDD construction process continues until primary outputs are reached. This approach, however, ignores the specific structure of a Boolean network. As a result, the elimination often stops at boundaries which are not natural to the specific Boolean network. This approach may also cause memory blow-up. The second approach is based on iterative elimination [10] which is quite similar to the eliminate procedure in SIS [19]. In the process, BDD node count is used as the cost function to guide the elimination.

To comply with the mainstream synthesis flow, an approach similar to that of $[19,10]$ has been adopted in it BDS. However, due to the efficiency of described techniques for BDD manipulations, our eliminate is orders of magnitude faster than [10].

A generic algorithm for iterative elimination is shown in Algorithm 1. In collect all Boolean nodes which are eligible for collapsing into its fanout are collected. In execute the Boolean network is modified to lock the recent changes. The BDD manager is reordered before the next collect. The whole process is iterated until no Boolean node can be eliminated.

```
candidate = collect(bddmgr,network);
while(candidate) {
    execute(bddmgr,candidate);
    reorder(bddmgr);
    candidate = collect(bddmgr, network); }
```

Algorithm 1: Eliminate
In practice, a straightforward implementation of this process does not work. This is mainly because of the abuse of BDD variable reordering. When local BDDs are constructed for a Boolean network, an intermediate variable is created for each Boolean node. Therefore, in addition to all primary inputs, a BDD manager also contains all intermediate variables. The number of such variables could be huge even for a mediumsize circuits. Since the complexity of variable reordering could be exponential, reordering a BDD manager with large number of variables will severely degrade the overall runtime performance.

```
candidate = collect(bddmgr,network);
while(candidate) {
    execute (bddmgr, candidate);
    newbddmgr = bddMapping(bddmgr, network);
    reorder(newbddmgr);
    free (bddmgr);
    bddmgr = newbddmgr;
    candidate = collect(bddmgr,network); }
```

Algorithm 2: Efficient Iterative Eliminate Paradigm

In $B D S$, new BDD manipulation techniques are developed to make the approach feasible in practice. Algorithm 2 shows the modified version of Algorithm 1. In this algorithm, instead of using a single BDD manager for all the operations, a new BDD manager is initialized in each iteration. All BDDs after one iteration are mapped (to be explained later) into the new BDD manager. A variable reordering is performed for the new BDD manager. The old BDD manager is abandoned. The process is iterated until no Boolean nodes can be eliminated.

The need for a new BDD manager and BDD mapping technique can be justified as follows. A typical approach for variable reordering is based on adjacent variable swapping. To find the optimal position for a variable, a bulk of unique table will be traversed. During the process of eliminate, the removal of one Boolean node from the Boolean network corresponds to the demise of one variable in the BDD manager, so the variable will not be used again. Let us refer these variables to as unused variables. After the termination of one iteration many Boolean nodes have been removed, and the BDD manager contains large number of unused variables. Table 3 shows the reduction of Boolean nodes after first iteration. It can be found that about $63 \%$ variables in the BDD manager become unused. It is obvious that performing variable reordering in a BDD manager with large number of unused variables is very inefficient.

| Circuit | Before | After | Reduction |
| :--- | :---: | :---: | :---: |
| C1355 | 474 | 60 | $88 \%$ |
| C1908 | 325 | 94 | $72 \%$ |
| C2670 | 656 | 281 | $58 \%$ |
| C3540 | 793 | 344 | $57 \%$ |
| C432 | 123 | 63 | $49 \%$ |
| C499 | 162 | 57 | $65 \%$ |
| C5315 | 1228 | 387 | $69 \%$ |
| C6288 | 2338 | 704 | $70 \%$ |
| C7552 | 1829 | 455 | $76 \%$ |
| C880 | 296 | 122 | $59 \%$ |
| dalu | 764 | 241 | $69 \%$ |
| des | 681 | 294 | $57 \%$ |
| mult32 | 5507 | 2467 | $56 \%$ |
| pair | 818 | 450 | $45 \%$ |
| Total | 15994 | 6019 | $63 \%$ |

Table 3: Number of Boolean node reduction after first eliminate

Instead of reordering the BDD manager with large number of unused variables, a new BDD manager with set of used variables is initialized. BDDs are then transfered into the
new BDD manager using our bddPool mechanism (see Section 3.2). During the process, variables are substituted according to $\mathcal{M}$, where $\mathcal{M}$ is the mapping of variables between the old and new BDD managers. When all BDDs are reconstructed in the new BDD manager, a set of BDDs which are graphically isomorphic to the original ones, but much more compact in the range of indexes, are obtained. This process is referred to as a BDD mapping.

Table 4 shows the results of our iterative eliminate paradigm. The results of [10] are also listed for comparison. On average, our eliminate is 85 times faster than [10]. The runtime advantage of $B D S$ becomes stronger for larger test cases. Although BDS is targeting multi-level implementation, it is obvious that the iterative eliminate paradigm in $B D S$ will also be useful for PTL synthesis as well. Due to the efficient way of handling BDD variable reordering, our iterative eliminate paradigm has the capability to handle very large circuits.

| Circuit | Chaudhry et al $[10]$ |  | BDS |  |
| :--- | :---: | :---: | :---: | :---: |
|  | BDD nodes | CPU $(\mathrm{s})$ | BDD nodes | CPU $(\mathrm{s})$ |
| C1355 | 211 | 270 | 207 | 0.3 |
| C1908 | 310 | 25.4 | 276 | 0.6 |
| C2670 | 615 | 197 | 527 | 1.7 |
| C3540 | 974 | 101.7 | 901 | 3.2 |
| C432 | 181 | 4.5 | 183 | 0.5 |
| C499 | 196 | 2.4 | 228 | 0.2 |
| C5315 | 1008 | 307.6 | 918 | 4.0 |
| C6288 | 1677 | 540.7 | 1507 | 4.4 |
| C7552 | 1592 | 382.1 | 1227 | 6.4 |
| C880 | 298 | 7.5 | 300 | 0.4 |
| Total | 7066 | 1838.9 | 6274 | 21.7 |

Table 4: Results of iterative eliminate paradigm. Both experiments are carried out on Pentium-200.

## 5. EXPERIMENTAL RESULTS

The experiments have been conducted on a Pentium-III/500 machine running Linux. Most large combinational circuits in LGSynth91 test case suite are covered in the experiment. All test cases are synthesized by both BDS and SIS (script.rugged). The results are then mapped by the SIS mapper.

Table 5 shows the experimental results. Two testcases, dalu and vda, have been singled out from this table for special illustration. In summary, BDS uses about $3 \%$ more area than SIS. The delay of circuits synthesized by BDS is $13 \%$ smaller. The memory used by BDS is $30 \%$ less than SIS. It should be noted that the memory usages reported for SIS in Table 5 only include the memory used by logic optimization procedures in script.rugged. Memory used by full_simplify is not included. Since global BDDs are constructed during full_simplify, which is not the case for $B D S$, it is unfair to compare the total memory used by script.rugged with BDS. In terms of runtime, BDS demonstrates superior advantage over SIS, it is more than 8 times faster than SIS. We must mention that compared with real industrial circuits, all the test cases used in this experiment are very small. We run both BDS and SIS on a set of circuits with more than 40 K gates. BDS was over 100 times faster than SIS. Since those circuits are not in the public domain, the experimental results are not published

| Circuit | SIS |  |  |  | BDS-1.3 |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Area | Delay | CPU (s) | Mem (M) | Area | Delay | CPU (s) | Mem (M) |
| C1355 | 689 | 39.40 | 6.6 | 1.2 | 711 | 45.60 | 0.4 | 1.0 |
| C1908 | 695 | 68.60 | 8.1 | 1.2 | 730 | 65.00 | 0.8 | 1.0 |
| C3540 | 1695 | 81.40 | 16.1 | 3.3 | 1713 | 81.20 | 3.6 | 1.9 |
| C432 | 290 | 75.90 | 46.1 | 0.7 | 357 | 78.40 | 0.2 | 0.5 |
| C499 | 689 | 39.40 | 6.8 | 0.9 | 708 | 43.60 | 0.6 | 0.5 |
| C5315 | 2286 | 68.60 | 10.2 | 3.1 | 2402 | 70.50 | 5.3 | 3.0 |
| C6288 | 4631 | 237.8 | 21.8 | 4.1 | 4677 | 178.3 | 3.8 | 1.1 |
| C7552 | 3038 | 115.70 | 54.2 | 4.9 | 3112 | 83.30 | 4.2 | 4.8 |
| C880 | 567 | 56.10 | 1.9 | 1.0 | 563 | 43.20 | 0.7 | 0.8 |
| pair | 2274 | 74.30 | 16.1 | 2.5 | 2466 | 52.60 | 2.1 | 2.0 |
| rot | 965 | 51.60 | 4.5 | 2.0 | 1025 | 51.90 | 1.0 | 0.9 |
| Total | 17819 | 908.8 | 192.4 | 24.9 | 18464 | 793.6 | 22.7 | 17.5 |

Table 5: Comparison between BDS-1.3 and SIS. The memory reported for SIS does not include the memory used by full_simplify.
here. However, one conclusion we can reach is that the overall complexity of $B D S$ is lower than that of SIS.

### 5.1 Analysis

There are two causes which contribute to the larger circuit area obtained with BDS. First, BDS has a capability to perform XOR and MUX decompositions. XORs and MUXes have been represented explicitly on the factoring trees and in the final BLIF format generated by BDS. However, due to the weak capability to identify XORs and MUXes by the tree-based technology mapper, only a small fraction of XORs and MUXes synthesized by BDS can be mapped to XOR and MUX gates. The same problem has been observed in our previous experiment [18].

Second, currently BDS does not have the capability to perform don't care minimization. If the redundancy can not be removed by eliminate, it will most likely remain in the final synthesized circuits. Lack of such a capability is the major hold-back for the current version of $B D S$. Shown in Table 6 are the synthesis results for dalu and vda. The results can be greatly improved by applying full_simplify on the circuits synthesized by BDS. However, the area and delay of dalu is still $50 \%$ more than SIS. Extensive comparison between BDS and SIS should be done for dalu.

| Circuit | SIS |  | BDS-1.3 |  |  |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  | no full_simplify | full_simplify |  |  |
|  |  |  | Area | Delay | Area | Delay |
| dalu | 1307 | 58.40 | 2680 | 103.5 | 1927 | 93.3 |
| vda | 837 | 39.8 | 1380 | 32.6 | 1049 | 43.2 |

Table 6: The effect of full_simplify.

## 6. CONCLUSION

In this paper, a new BDD-based logic optimization system has been presented. The experimental results show that BDD-based logic optimization is a promising alternative to the existing logic optimization approaches. Currently, $B D S$ is still a very dynamic program. We anticipate the performance of $B D S$ will be promoted to a higher level in the near future. A preliminary version of $B D S$ can be downloaded from [20].

Compared with the state-of-the-art logic synthesis methodology, BDD-based logic synthesis is new but much less mature.

It is too soon to conclude that it will be a practical alternative to the widely accepted methodology. Extensively fundamental research has to be done to make this approach a truly successful synthesis method. We hope our research can initiate a new round of research in logic synthesis area in the years to come.

## Acknowledgment:

This work has been supported by a grant from NSF under contract No. MIP-9613864.

## References

[1] R. Rudell, "Tutorial : Design of a logic synthesis system," in Proc. 33rd Design Automation Conference, 1996, pp. 191-196.
[2] S. B. Akers, "Functional Testing with Binary Decision Diagrams," in Eighth Annual Conference on FaultTolerant Computing, 1978, pp. 75-82.
[3] Randal E. Bryant, "Graph-Based Algorithms for Boolean Function Manipulation," IEEE Trans. on Computer, vol. 35, no. 8, pp. 677-691, August 1986.
[4] K. Brace, R. Rudell, and R. Bryant, "Efficient Implementation of a BDD Package," in Proc. Design Automation Conference, 1990, pp. 40-45.
[5] Yung-Te Lai, Kuo-Rueih Pan, and Massoud Pedram, "OBDD-Based Function Decomposition: Algorithms and Implementattion," IEEE Trans. on CAD, vol. 15, no. 8, pp. 977-990, August 1996.
[6] Shih-Chieh Chang, M. Marek-Sadowska, and T. Hwang, "Technology Mapping for TLU FPGA's Based on Decomposition of Binary Decision Diagrams," IEEE Trans. on CAD, vol. 15, no. 10, pp. 1226-1235 October 1996.
[7] Jason Cong and Yuzheng Ding, "Beyond the Combinational Limit in Depth Minimization for LUTBased FPGA Designs," in IEEE International Conference on Computer-Aided Design, 1993, pp. 110-114.
[8] K. Yano, Y. Sasaki, K. Rikino, and K. Seki, "Top-Down Pass Transistor Logic Design," IEEE J. Solid-State Circuits, vol. 31, no. 6, pp. 792-803, June 1996.
[9] P. Buch, A. Narayan, R. Newton, and A. Sangiovanni-Vincentelli, "On Synthesizing Pass Transistor Logic," in Intl. Workshop on Logic Synthesis, 1997.
[10] R. Chaudhry, T. Liu, A. Aziz, and J. Burns, "Area-Oriented Synthesis for Pass-Transistor Logic," in International Conference on Computer Design, 1998, pp. 160-167.
[11] Kevin Karplus, "Using if-then-else DAGs for Multi-Level Logic Minimization," Tech. Rep. UCSC-CRL-88-29, University of California Santa Cruz, 1988.
[12] V. Bertacco and M. Damiani, "The Disjunctive Decomposition of Logic Functions," in IEEE International Conference on Computer-Aided Design, 1997, pp. 78-82.
[13] Ted Stanion and Carl Sechen, "Boolean Division and Factorization Using Binary Decision Diagrams," IEEE Trans. on CAD, vol. 13, no. 9, pp. 1179-1184, September 1994.
[14] S. Minato, "Fast Factorization Method for Implicit Cube Set Representation," IEEE Trans. on CAD, vol. 15, no. 4, pp. 377-384, April 1996.
[15] M. A. Thornton and V. S. S. Nair, "Behavioral Synthesis of Combinational Logic Using Spectral Based Heuristics," ACM Transactions on Design Automation of Electronic Systems, vol. 4, no. 2, pp. 219-230, April 1999.
[16] S. Yamashita, K. Yano, Y. Sasaki, Y. Akita, H Chikata, K. Rikino, and K Seki, "Pass-Transistor/CMOS Collaborated Logic: The Best of Both Worlds," in Symposium on VLSI Circuits Digest of Technical Papers, 1997, pp. 31-32.
[17] C. Yang and M. Ciesielski, "Synthesis for Mixed CMOS/PTL Logic: Preliminary Results," in International Workshop on Logic Synthesis, 1999.
[18] C. Yang, V. Singhal, and M. Ciesielski, "BDD Decomposition for Efficient Logic Synthesis," in International Conference on Computer Design, 1999, pp. 626-631.
[19] E. Sentovich et al., "SIS: A System for Sequential Circuit Synthesis," Tech. Rep. UCB/ERL M92/41, ERL, Dept. of EECS, Univ. of California, Berkeley., 1992.
[20] http://www.ecs.umass.edu/ece/labs/vlsicad/ciesielski.html.

