# Verification of Scheduling in High-level Synthesis 

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#### Abstract

This paper describes a formal method for checking the equivalence between two descriptions of the target system, one before and the other after scheduling. The descriptions are represented as finite state machines with data paths (FSMD). The basic principle is to show that any computation of one FSMD is covered by a computation on the other, a computation being characterized by a concatenation of paths in the FSMD. These notions are formalized in the paper. The method is strong enough to accommodate merging of the segments in the original behaviour by the typical scheduler such as DLS, a feature common in scheduling. The method also works for limited arithmetic transformations. Although the proposed method is found to have a non-polynomial worst case complexity, many non-trivial examples encounter a low polynomial order of complexity. The technique is illustrated with an example.


## 1 Introduction

High-level synthesis is the process of generating the register transfer level (RTL) design from the behavioural description. The synthesis process consists of several interdependent sub-tasks such as, specification, compilation, scheduling, allocation and binding. The operations in the behavioural description are assigned time steps through the scheduling process. Input to the scheduling phase is a control data flow graph (CDFG)[3]. While a CDFG is better suited to scheduling algorithms, an FSMD is a more appropriate model for verification. We therefore construct FSMDs from the CDFGs before and after scheduling. In the process of scheduling, operations are often moved across basic block boundaries for various optimizations. In general several transformations may be made to improve the performance of a design. For example, path based scheduling techniques [7] perform several such non-trivial path based transformations. Hence, it is important to ensure that the
scheduling process preserves the behaviour of the original specification, irrespective of the scheduling technique that is used. The objective of this work is to check that the design descriptions before and after scheduling, as represented by FSMDs, are computationally equivalent.

The equivalence problem of FSMDs (EPFSMD) is the same as the equivalence problem of flowchart schemas $[4,6]$ which is undecidable and not even partially decidable[6]. However, since the final targeted hardware has only a finite datapath, the restricted problem can be reduced to the equivalence problem of FSM models (EPFSM) which is decidable. Unfortunately, an FSMD with an n-bit datapath results in a number of states of the order of $2^{k n}$, where k is the number of storage elements of $n$ bits. The value of $k n$ easily exceeds several hundreds. Thus, deciding EPFSMD with a finite datapath by reducing them to EPFSM is of little use in practice. On the other hand specialized analytical treatments, such as the work described here, may aid in revealing problems in the working of the algorithm which may never use the finiteness in producing the output which is to be checked. In this case the equivalence checking algorithm would identify paths that are not matched up, which could be particularly helpful in fixing the scheduling algorithm. This benefit would normally be lost by trying to reduce a finite EPFSMD to EPFSM.

Most of the algorithms proposed in the literature can successfully verify the basic block based scheduling but apparently fail to verify when structure of the scheduled FSMD differs from the input FSMD due to path based transformation. In this paper, we propose a scheduling verification method which is strong enough to work even when the basic path structure is changed by the scheduler. This method formally establishes equivalence between the FSMDs before and after scheduling.

This paper is organized as follows. In section 2, FSMDs and the notions of computations on FSMDs and the equivalence of FSMDs are defined. The verification method is described in section 3 . the complexity of the proposed method is treated in section 4. An example has been treated in sec-
tion 5 to illustrate the working of the algorithm. Some experimental results have been given in section 6 . The paper is concluded in section 7 .

## 2 FSMDs and their Equivalence

### 2.1 FSMDs

An FSMD (finite state machine with data-path) is a universal specification model, proposed by Gajski in [2], that can represent all hardware designs. The model is used in the present work with the addition of a reset state, for encoding the designs to be verified. The FSMD is defined as an ordered tuple $\left\langle Q, q_{0}, I, V, O, f, h\right\rangle$, where

1. $Q=\left\{q_{0}, q_{1}, q_{2}, \ldots q_{n}\right\}$ is the finite set of control states,
2. $q_{0} \in Q$ is the reset state,
3. $I$ is the set of primary input signals and $\Sigma_{I}$ is the input alphabet,
4. $V$ is the set of storage variables and $\Sigma$ is the set of all data storage states or simply, data states,
5. $O$ is the set of primary output signals and $\Sigma_{O}$ is the output alphabet,
6. $f: Q \times S \rightarrow Q$, is the state transition function and
7. $h: Q \times S \rightarrow U$, is the update function of the output and the storage variables, where $U$ and $S$ are as defined below.
(a) $U=\{x \Leftarrow e \mid x \in O \cup V$ and $e \in E\}$ represents a set of storage or output assignments, from variables (storage or output) or expressions constructed over (input or storage) variables. Thus, $E=\{g(x, y, z, \ldots) \mid x, y, z, \ldots \in I \cup V\}$ represents a set of arithmetic expressions over the set $I \cup V$.
(b) $S=\{R(a, b) \mid a, b \in E$ and $R$ is any arithmetic relation $\}$ represents a set of status signals as a result of comparisons $(=, \neq,>, \geq,<, \leq)$ between two expressions from the set $E$.

It may be noted that we have not introduced final states in the FSMD model as we assume that the systems work in an infinite outer loop.

### 2.2 Walks and Transformations along a Walk

A (finite) walk $\alpha$ from $q_{i}$ to $q_{j}$, where $q_{i}, q_{j} \in Q$, is a finite transition sequence of states of the form $\left\langle q_{i}=q_{1} \overrightarrow{c_{1}} q_{2} \overrightarrow{c_{2}}\right.$, $\left.\cdots, \overrightarrow{c_{n-1}} q_{n}=q_{j}\right\rangle$ such that $\forall l, 1 \leq l \leq n-1, \exists c_{l} \in S$ such that $f\left(q_{l}, c_{l}\right)=q_{l+1}$, and $q_{k}, 1 \leq k \leq n-1$, are all distinct.

The state $q_{n}$ may be identical to $q_{1}$. The condition of execution of the walk $\alpha=\left\langle\begin{array}{llllll}q_{l_{0}} & \overrightarrow{c_{0}} & q_{l_{1}} & \overrightarrow{c_{1}} & q_{l_{2}} & \cdots \overrightarrow{c_{k-1}}\end{array} q_{l_{k}}\right\rangle, R_{\alpha}$, is a logical expression over the variables in $V$ such that $R_{\alpha}$ is satisfied by the (initial) data state at $q_{l_{0}}$ iff the walk $\alpha$ is traversed.

We assume that inputs and outputs occur through named ports. The $i^{\text {th }}$ input from port $P$ is a value represented as $P_{i}$. Thus if some variable $v$ stores input from port $P$ (for the $i^{\text {th }}$ time along a walk), it is equivalent to the assignment $v \Leftarrow P_{i}$.

The simple data transformation of a walk $\alpha$ over $V\left(s_{\alpha}\right)$ : It is an ordered tuple $\left\langle e_{i}\right\rangle$ of algebraic expressions over the variables in $V$ and the inputs in $I$ such that the expression $e_{i}$ represents the value of the variable $v_{i}$ after the execution of the walk in terms of the initial data state (i.e., the values of the variables at the initial control state) of the walk.

Taking into account outputs that may occur in a walk, the data transformation $r_{\alpha}$ of a walk $\alpha$ over $V$ is the tuple $\left\langle s_{\alpha}, O_{\alpha}\right\rangle$, where the output list $O_{\alpha}=\operatorname{OUT}\left(P_{i_{1}}, e_{1}\right)$, $\left.\operatorname{OUT}\left(P_{i_{2}}, e_{2}\right), \ldots\right]$. For every expression $e$ output to port $P$ along the walk $\alpha$, there is an $\operatorname{OUT}(P, e)$ in the list, in the order in which the outputs occurred.


Figure 1. $M_{0}$, the FSMD of GCD before scheduling

Computation of the condition of execution $R_{\alpha}$ can be by backward substitution or by forward substitution. The former is more readily described and is based on the following rule: If a predicate $c(y)$ is true after execution of $y \leftarrow g(y)$, then the predicate $c(g(y))$ must have been true before the execution of the statement [6]. The transformation $s_{\alpha}$ is found indirectly using the same principle. The forward substitution method of finding $R_{\alpha}$ is based on symbolic execution.

### 2.3 Characterization of Walks and their Concatenations

The characteristic formula $\tau_{\alpha}$ of a walk $\alpha$ with initial storage and input variables as $\bar{v}$, final variables as $\bar{v}_{f}$ and outputs along the walk as $O$ is $\tau_{\alpha}\left(\bar{v}, \bar{v}_{f}, O\right)=R_{\alpha}(\bar{v}) \wedge\left(\bar{v}_{f}=\right.$ $\left.s_{\alpha}(\bar{v})\right) \wedge\left(O=O_{\alpha}(\bar{v})\right)$, where $s_{\alpha}$ is the data transformation and $O_{\alpha}$ output list in the walk $\alpha$.

Let $\tau_{\alpha}\left(\bar{v}, \bar{v}_{f}, O\right): R_{\alpha}(\bar{v}) \wedge\left(\bar{v}_{f}=s_{\alpha}(\bar{v})\right) \wedge\left(O=O_{\alpha}(\bar{v})\right)$ be the characteristic formula of the walk $\alpha$ and $\tau_{\beta}\left(\bar{v}, \bar{v}_{f}, O\right): R_{\beta}(\bar{v}) \wedge\left(\bar{v}_{f}=s_{\beta}(\bar{v})\right) \wedge\left(O=O_{\beta}(\bar{v})\right)$ be the characteristic formula of the walk $\beta$. The characteristic formula for the concatenated walk $\alpha \beta \quad$ is $\quad \tau_{\alpha \beta}\left(\bar{v}, \bar{v}_{f}, O\right)=\exists \bar{v}_{\alpha} \exists O_{1} \exists O_{2}\left(\tau_{\alpha}\left(\bar{v}, \bar{v}_{\alpha}, O_{1}\right) \wedge\right.$ $\left.\tau_{\beta}\left(\bar{v}_{\alpha}, \bar{v}_{f}, O_{2}\right)\right)=R_{\alpha}(\bar{v}) \wedge R_{\beta}\left(s_{\alpha}(\bar{v})\right) \wedge\left(\bar{v}_{f}=s_{\beta}\left(s_{\alpha}(\bar{v})\right)\right) \wedge$ $\left(O=O_{\alpha}(\bar{v}) O_{\beta}\left(s_{\alpha}(\bar{v})\right)\right) . O$ is the concatenated output list of $O_{\alpha}(\bar{v})$ and $O_{\beta}\left(s_{\alpha}(\bar{v})\right)$. The detail of incrementing the input indices on each port in the formulas for $\beta$ to start after the last index of the corresponding port in $\alpha$ has been omitted for notational clarity.

### 2.4 Computations on FSMDs and their Path Covers

A computation of an FSMD is a finite walk from the reset state $q_{0}$ back to itself without having any intermediary occurrence of $q_{0}$ (as a new computation starts from the reset state). A computation $c$ of an FSMD $M$ may be characterized as $\tau_{c}\left(\bar{v}_{i}, \bar{v}_{f}, O\right): R_{c}\left(\bar{v}_{i}\right) \wedge\left(\bar{v}_{f}=s_{c}\left(\bar{v}_{i}\right)\right) \wedge\left(O=O_{c}\left(\bar{v}_{i}\right)\right)$, where $\bar{v}_{i}$ is the vector of initial input and data state with which the computation is started, $R_{c}$ is a satisfiable condition over the domain of $I$ and $V, s_{c}$ is a function over this domain to the co-domain of values over $V$ and $O_{c}$ is the concatenation of the output lists resulting from output operations along $c$.

Two computations $c_{1}$ and $c_{2}$ having the characteristic formulae $\tau_{c_{1}}$ and $\tau_{c_{2}}$, respectively, are said to be equivalent if $R_{c_{1}}=R_{c_{2}}, r_{c_{1}}=r_{c_{2}}$. The computational equivalence of two walks $p_{1}$ and $p_{2}$ is denoted as $p_{1} \simeq p_{2}$. Equivalence checking of walks, therefore, consists in establishing the computational equivalence of the respective conditions of execution and the respective data transformations.

A finite set of paths ${ }^{1} P=\left\{p_{0}, p_{1}, p_{2}, \ldots, p_{k}\right\}$ is said to cover an FSMD $M$ if any computation $c$ of $M$ can be looked upon as a concatenation of paths from $P . P$ is said to be a finite path cover of the FSMD M.

[^0]
### 2.5 Arithmetic Expressions and their Normalization

Since the condition of execution and the data transformation of a walk involve the whole of integer arithmetic, checking equivalence of walks reduces to the validity problem of first order logic; the latter is undecidable because a canonical form does not exist for integer arithmetic. Instead, in this work we use the following normal form adapted from [5, 8].

Every formula is converted into the conjunctive normal form; every conjunct, therefore, is a disjunction of literals where a literal is an atomic formula (atom) or its negation. An atom is a boolean variable or an arithmetic relation of the form $S r 0$, where $S$ is a normalized sum, $r \in\{\leq, \geq,=, \neq\}$. The relation $>(<)$ can be reduced to $\geq$ $(\leq)$ over integers. A normalized sum is a sum of terms with at least one constant term; each term is a product of primaries with a non-zero constant primary; each primary is a storage variable, an input variable or an output variable or of the form $\operatorname{abs}(s), \bmod \left(s_{1}, s_{2}\right), \exp \left(s_{1} . s_{2}\right)$ or $\operatorname{div}\left(s_{1}, s_{2}\right)$, where $s, s_{1}$, and $s_{2}$ are normalized sums. Any normalized sum is arranged by lexicographic ordering of its constituent subexpressions from the bottom-most level. The common subexpressions in a sum are collected. Thus, $x^{2}+3 x+4 z+7 x$ is simplified to $x^{2}+10 x+4 z+0$. A relational literal is reduced by a common constant factor, if any, and the literal is accordingly simplified. For example, $3 x^{2}+9 x y+6 z+7 \geq 0$ is simplified to $x^{2}+3 x y+2 z+2 \geq 0$, where $\lfloor 7 / 3\rfloor=2$. A conjunct $C=l_{1} \vee l_{2} \vee \ldots \vee l_{n}$ is first expressed as $\neg\left(\neg l_{1} \wedge \neg l_{2} \wedge \ldots \wedge \neg l_{n}\right)$ and then literals are deleted by the rule "if $\left(l \Rightarrow l^{\prime}\right)$ then $l \wedge l^{\prime} \equiv l . " C$ reduces to true if $\neg l_{i} \Rightarrow l_{j}$ for $1 \leq i, j \leq n$. Symmetry of $\{=, \neq\}$, reflexivity of $\{\leq, \geq,=\}$ and irreflexivity of $\{\neq\}$ are accounted for by the above transformations. The above normal form may be shown to be canonical for multivariate polynomials.

### 2.6 Equivalence of FSMDs

Let $M_{0}$ be the FSMD representation of the CDFG given as the input to the scheduler and $M_{1}$ be the FSMD of the scheduled behaviour. Our main goal is to verify whether $M_{0}$ behaves exactly as $M_{1}$. This means that for all possible input sequences, $M_{0}$ and $M_{1}$ produce the same sequences of output values and eventually, when the respective reset states are re-visited, they are visited with the same storage element values. In other words, for every computation from the reset state back to itself of one FSMD, there exists an equivalent computation from the reset state back to itself in the other FSMD and vice-versa.

Thus two FSMDs $M_{0}$ and $M_{1}$ are said to be computationally equivalent if for any computation $c_{0}$ of $M_{0}$, there exists a computation $c_{1}$ of $M_{1}$ such that $c_{0}$ and $c_{1}$ are computationally equivalent and vice-versa.

The following theorem, stated without proof, is key to our algorithm for checking the equivalence of two FSMDs.

Theorem 1 Two FSMDs $M_{0}$ and $M_{1}$ are computationally equivalent if there exists a finite cover $P_{0}=$ $\left\{p_{00}, p_{01}, \ldots, p_{0 l}\right\}$ of $M_{0}$ for which there exists a set $P_{1}^{0}=$ $\left\{p_{10}^{0}, p_{11}^{0}, \ldots, p_{1 l}^{0}\right\}$ of paths of $M_{1}$ such that $p_{0 i} \simeq p_{1 i}^{0}$, $0 \leq i \leq l$ and vice-versa.

The following (inductive) notion of corresponding states will be used in the algorithm to be presented. Let $M_{0}=$ $\left\langle Q_{0}, q_{00}, I, V_{0}, O, f_{0}, h_{0}\right\rangle$ and $M_{1}=\left\langle Q_{1}, q_{10}, I, V_{1}, O, f_{1}, h_{1}\right\rangle$ be the two FSMDs having identical input and output sets, $I$ and $O$, respectively, and $q_{0 i}, q_{0 k} \in Q_{0}$ and $q_{1 j}, q_{1 l} \in Q_{1}$.

- The respective reset states $q_{00}, q_{10}$ are corresponding states.
- If $q_{0 i} \in Q_{0}$ and $q_{1 j} \in Q_{1}$ are corresponding states and there exist $q_{0 k} \in Q_{0}$ and $q_{1 l} \in Q_{1}$ such that, for some path $\alpha$ from $q_{0 i}$ to $q_{0 k}$ in $M_{0}$, there exists a path $\beta$ from $q_{1 j}$ to $q_{1 l}$ in $M_{1}$ such that $\alpha \simeq \beta$, then $q_{0 k}$ and $q_{1 l}$ are corresponding states.


## 3 Verification Method

The above theorem, therefore, suggests a verification method which consists of the following steps:

1. Construct the set $P_{0}$ of paths of $M_{0}$ so that $P_{0}$ covers $M_{0}$. Let $P_{0}=\left\{p_{00}, p_{01}, \cdots, p_{0 k}\right\}$.
2. Show that $\forall p_{0 i} \in P_{0}$, there exists a path $p_{1 j}$ of $M_{1}$ such that $p_{0 i} \simeq p_{1 j}$.
3. Repeat steps 1 and 2 with $M_{0}$ and $M_{1}$ interchanged.

Owing to the presence of loops it is difficult to find a path cover of the whole computation comprising only finite paths. So any computation is split into paths by putting cutpoints at various places in the FSMD so that each loop is cut in at least one cutpoint. The set of all paths from a cutpoint to another cutpoint without having any intermediary cutpoint is a path cover of the FSMD. The method of decomposing an FSMD by putting cutpoints is identical to the Floyd-Hoare's method of program verification [1, 5]. We choose the cutpoints in any FSMD as follows.

1. The reset state is chosen.
2. Any state with more than one outward transitions is also chosen.

Obviously, cutpoints chosen by the above rules cut each loop of the FSMD in at least one cutpoint, because each
internal loop has an exit point (ensured by our notion of computation in §2).

In the following we propose one method which combines the first two steps listed above into one. More specifically, the method constructs a path cover of $M_{0}$ and also finds its equivalent path set in $M_{1}$ hand-in-hand. An initial set of cutpoints is chosen for $M_{0}$ as described above. The reset state of $M_{0}$ is always a cutpoint of $M_{0}$ and the reset states of $M_{0}$ and $M_{1}$ is the initial pair of corresponding states. Starting from a corresponding state $q_{0 i}$ of $M_{0}$ the algorithm traverses all the paths leading out of $q_{0 i}$ to the next cutpoint in $M_{0}$ and for each path it tries to find a corresponding equivalent path in $M_{1}$. On success the end points of the two paths may be recorded as a new pair of corresponding points. Otherwise, the path in $M_{0}$ is extended in all possible ways (without re-entering loops) and again matching paths are sought in $M_{1}$ for each extension. When all possibilities are exhausted without finding a match the algorithm reports a failure. The algorithm continues until all pairs of corresponding points are processed. The following pseudocode describes this process more precisely.

### 3.1 Verification Algorithm

Step 1: Insert cutpoints in $M_{0}$ by the following rules.

- the start state is a cutpoint,
- any state with more than one outward transition is a cutpoint.


## Step 2:

/*Main data stores:
$\eta$ : Set of corresponding nodes
$P_{0}$ : path cover of $M_{0}$
$P_{1}^{0}$ : paths in $M_{1}$ with matching paths in $P_{0}$
Working data stores:
F: list of paths of $M_{0}$ starting with nodes having corresponding nodes but ending with nodes whose corresponding nodes have not yet been found
P : Working list of corresponding nodes from which paths will be examined */
$\mathrm{F}:=[] ; \mathrm{P}_{0}:=[] ; \mathrm{P}_{1}{ }^{0}:=[]$;
$\eta:=\left\{\left\langle q_{00}, q_{10}\right\rangle\right\} ;$
$\mathrm{P}:=\left\{\left\langle q_{00}, q_{10}\right\rangle\right\}$;
while ( P is not empty $\| \mathrm{F}$ is not empty )
\{ // main loop continues till termination
if ( F is empty )
// new paths starting from entries in P to be examined
$\left\{\left\langle q_{0 i}, q_{1 j}\right\rangle:=\right.$ deQ P ;
put in F all the paths from $q_{0 i}$ to its successor cutpoints (in $M_{0}$ ) ;
\} else // now work on the un-matched path frontier

```
    { \beta:= deQ F ; // endPtNd ( }\beta\mathrm{ ) is un-matched!
        if (( \alpha= findEquivalentPath ( }\beta,\mp@subsup{q}{1j}{}))!=\mathrm{ NULL )
        { if (!\langleendPtNd ( }\beta)\mathrm{ , endPtNd ( }\alpha)\rangle\in\eta
            enQ ( P, \langle endPtNd ( }\beta)\mathrm{ , endPtNd ( }\alpha)\rangle)
            // new paths will start from here
        \eta:= \eta\cup{\langle endPtNd( }\beta)\mathrm{ , endPtNd ( }\alpha)\rangle
        P
    } else // no match
    { // so continue along all paths through successors
        if ( the path is marked NOT_EXTENDIBLE ) fail ;
        tF := all the paths obtained by concatenating to }
        all the paths from endPtNd ( }\beta)\mathrm{ to all the successor
        cutpoints of endPtNd ( }\beta\mathrm{ );
        if ( endPtNd of any member of tF is a node of the
            same path other than its start node )
            fail;
        if ( endPtNd of any member of tF is same
            as its start node | the reset state)
            mark the path as NOT EXTENDIBLE ;
        F:= append (tF,F );
    } // else-if
    } // else-if
} // end while
```

Step 3: Identify the cutpoints in $M_{1}$
Step 4: Repeat the same procedure as described in Step 2 with the roles of $M_{0}$ and $M_{1}$ interchanged.

Step 5: If it succeeds for both Step 2 and Step 4 then report $M_{0}$ and $M_{1}$ are computationally equivalent. Otherwise report a failure.

The functions used are specified as follows.

- findEquivalentPath $\left(\beta, q_{1 j}\right)$ : It tries to find a path $\alpha$ in $M_{1}$ so that $R_{\alpha}=R_{\beta}$ and $r_{\alpha}=r_{\beta}$. If such an $\alpha$ exist then this function returns $\alpha$, otherwise a NULL path.
- endPtNd $(\beta)$ : returns the state where the path $\beta$ terminates.


## 4 Algorithm Complexity

The complexity of the algorithm is determined in step 2. Let there be up to $n$ control states (nodes) in $M_{0}$ or $M_{1}$. This is an upper bound on the number of cutpoints in $M_{0}$ or $M_{1}$. The complexity of the function findEquivalent $\left(\beta, q_{1 j}\right)$ is proportional to the number of paths from $q_{1 j}$ to be examined, for a given $\beta$. If there are up to $k$ parallel edges between any two states, then up to $O(k n)$ paths of length $O(n)$ need to be examined. We avoid the $k^{n}$ possibilities of branching here because the paths are examined with respect to $\beta$; only the $k$ possibilities at each node of the path need to
be examined for only a single viable choice. Thus the time complexity of findEquivalent is $O\left(k n^{2}\right)$. This is not a tight upper bound, so we also consider the lower bound which is $\Omega(1)$. This represents the case when we obtain straight matches without any exploration.

While finding the equivalent paths in $M_{1}$ corresponding to those in $M_{0}$, in the best case we only need to consider a path from one cutpoint to the next. However, path extensions may be required. The maximum length of any path in $M_{0}$ after repeated path extensions is $O(n)$. Starting from a cutpoint, considering repeated path extensions and an edge multiplicity of $k$, up to $O\left(k^{n-1}\right)$, if $k \neq 1$ or $O\left(n^{2}\right)$, if $k=1$, paths of $M_{0}$ may need to be examined. Note that if $k \neq 1$, then along a path of $n$ nodes there are $k$ path segments of length one, $k^{2}$ path segments of length two, etc. The total number of such paths is $k+k^{2}+k^{3}+\ldots+k^{n-1}=k \frac{k^{n-1}-1}{k-1}$, which is $O\left(k^{n-1}\right)$. Since paths may start from any cutpoint, $O\left(n k^{n-1}\right)$, if $k \neq 1$ or $O\left(n^{3}\right)$, if $k=1$, paths of $M_{0}$ may need to be examined. The total number of pairs of the form $\left\langle\beta, q_{1 j}\right\rangle$ that need to be examined is $O\left(n^{2} k^{n-1}\right)$, if $k \neq 1$ or $O\left(n^{4}\right)$, if $k=1$. Thus, the overall worst case complexity of step 2 is $O\left(n^{4} k^{n}\right)$ if $k \neq 1$ or $O\left(n^{6}\right)$, if $k=1$ and $\Omega(n)$. In practice this is often not hit.

## 5 An Example



Figure 2. $M_{1}$, the FSMD of GCD after scheduling

The flow of step 2 of the proposed algorithm applied to the GCD example (fig 1 and fig 2) is briefly discussed here. The algorithm first considers the path $\left\langle q_{00} \rightarrow q_{01}\right\rangle$. The findEquivalentPath function successfully finds an equivalent path which is $\left\langle q_{10} \rightarrow q_{11}\right\rangle$. So, $\left\langle q_{01}, q_{11}\right\rangle$ are the corresponding pair of states. The path $\left\langle q_{01} \rightarrow q_{0 e} \rightarrow q_{00}\right\rangle$ has to be considered next. Its equivalent path is $\left\langle q_{11} \rightarrow\right.$
$\left.q_{12} \rightarrow q_{1 e} \rightarrow q_{10}\right\rangle$. For the path $\left\langle q_{01} \rightarrow q_{02}\right\rangle$, there is no equivalent path in $M_{1}$. So, this path needs to be concatenated with its successor paths. These concatenated paths are $\beta_{1}=\left\langle q_{01} \rightarrow q_{02} \rightarrow q_{03}\right\rangle$ and $\left.\beta_{2}=\left\langle q_{01} \rightarrow q_{02} \rightarrow q_{04}\right\rangle\right\}$. $\beta_{1}$ is again concatenated with its successor paths as it has no equivalent path in $M_{1}$. The concatenated paths have the same sequence of states $\left(q_{01} \rightarrow q_{02} \rightarrow q_{03} \rightarrow q_{01}\right)$ with different conditions of execution and data transformations. Similarly, $\beta_{2}$ is also concatenated with its successor paths. These concatenated paths are $\beta_{21}=\left\langle q_{01} \rightarrow q_{02} \rightarrow q_{04} \rightarrow\right.$ $\left.\left.q_{05}\right\rangle\right\}$ and $\left.\beta_{22}=\left\langle q_{01} \rightarrow q_{02} \rightarrow q_{04} \rightarrow q_{01}\right\rangle\right\}$. Path $\beta_{21}$ will be concatenated with two of its successor paths next. Now each of these concatenated paths has an equivalent path in $M_{1}$. These will be explored one by one in the following iterations. It may be noted that Step 2 takes only 18 iterations which is much better than the worst case complexity of $6^{6}=46656$ (here $k=1, n=6$ ).

## 6 Experimental Results

The proposed algorithm has been implemented in ' C ' and has been run for some standard high-level synthesis benchmarks as shown in table 1. These have been run on an Intel Pentium 4, 1.70 MHz, 256MB RAM machine. The number of states, number of paths explored in each FSMD $M_{0}$ and $M_{1}$, number of consecutive path segments merged by the scheduler and the CPU time are tabulated for each benchmark example. It is evident from table that execution time is sensitive on number of paths explored. It also may be noted from the table that run time of this algorithm is less sensitive on the number of states in the FSMDs. For example, in table 1, the run times of EWF and DCT are small compared to GCD and MODN even though EWF and DCT have greater number of states. These examples also suggest that the upper bound is not necessarily hit for practical scheduling verification cases.

## 7 Conclusions

Advances in VLSI technology have enabled its deployment into complex circuits. Synthesis flow of such circuits comprises various phases where each phase performs the task algorithmically providing for ingenious interventions of experts. The gap between the original behaviour and the finally synthesized circuits is too wide to be analyzed by any reasoning mechanism. The validation tasks, therefore, must be planned to go hand in hand with each phase of synthesis. The present work concerns itself with the validation of the scheduling phase. Both the behaviours prior to and after scheduling have been modeled as FSMDs. The validation task has been treated as an equivalence problem of FSMDs.

The method presented in this paper has been proved to be sound, completeness being ruled out by the fact that the

| Name | \#state in <br> FSMD |  | \#path <br> cover |  | \#path <br> extn | CPU <br> time <br> in ms |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $M_{0}$ | $M_{1}$ | $M_{0}$ | $M_{1}$ |  |  |
| DIFFEQ | 4 | 12 | 3 | 3 | 0 | 2.442 |
| EWF | 4 | 35 | 1 | 1 | 0 | 1.820 |
| GCD | 7 | 4 | 11 | 7 | 3 | 3.976 |
| DCT | 3 | 29 | 1 | 1 | 0 | 1.754 |
| TLC | 7 | 8 | 13 | 14 | 2 | 4.196 |
| MODN | 6 | 7 | 8 | 12 | 2 | 4.324 |
| PERFECT | 9 | 6 | 7 | 5 | 2 | 4.028 |

Table 1. Results for different high-level synthesis benchmarks
equivalence problem of FSMDs has been reported to be not even partially decidable. The method is strong enough to accommodate merging of the segments in the original behaviour by the typical scheduler such as, DLS [7]. It is also able to handle arithmetic transformations and expected to handle simple code motion. Similar methods reported in the literature have been found to fail under such situations. Although the proposed method is found to have a non-polynomial worst case complexity primarily because of presence of parallel edges between the same pair of states, the best case complexity is found to be linear. The initial experiments show that the algorithm is usable for practical equivalence checking cases of scheduling.

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[^0]:    ${ }^{1}$ A path is a walk in which all the states (nodes) are distinct. A cycle is like a path where the first and last nodes are identical but all other nodes are distinct. Here we allow our paths to be cycles also.

