Multi-level clustering for clock skew optimization

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ABSTRACT

Clock skew scheduling has been effectively used to reduce the clock period of sequential circuits. However, this technique may become impractical if a different skew must be applied for each memory element. This paper presents a new technique for clock skew scheduling constrained by the number of skew domains. The technique is based on a multi-level clustering approach that progressively groups flip-flops with skew affinity. This new technique has been compared with previous work, showing the efficiency in the obtained performance and computational cost. As an example, the skews for an OpenSparc with almost 16K flip-flops and 500K paths have been calculated in less than 5 minutes when using only 2 to 5 skew domains.

1. INTRODUCTION

The clock signal is distributed via wires and gates forming trees, meshes, or other structures. The clock skew is defined as the clock arrival time difference between memory elements. Traditionally, clock skew has been considered to be a problem, with negative impact on the clock period.

Nowadays, clock skew scheduling [1] is becoming a common practice in EDA flows. The intentional skew applied to the memory elements of a circuit can reduce the cycle period by borrowing time from the paths with the largest slacks and using it in the paths with the smallest slacks.

This optimization implies a certain degree of *wavepipelining* in the long paths that temporarily carry two waves of data on the same cycle. These paths must guarantee a minimum delay sufficient to safely separate the time distance between the two travelling waves.

An optimal solution for skew scheduling may potentially require a different skew for each memory element in the circuit. This may generate an excessive demand of buffers with an unacceptable overhead in area and power consumption. Skew optimization is usually constrained to a limited number of skew domains. Within each domain, the memory elements are assumed to have negligible skew.

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The correctness of skew optimization is finally determined by the *setup* and *hold* constraints that must be met for any combinational path between memory elements.

After clock skew scheduling, the hold constraints can be met by adding delays in the shortest paths [2, 3]. However, aggressive skews may result in an excessive amount of required delays with the corresponding impact in area and power.

1.1 Asynchronous circuits

Clock skew scheduling may also play an important role in asynchronous circuits. Recently, a new technique called *desynchronization* [4] has appeared as an automated process to transform synchronous circuits into asynchronous. This process replaces the clock of the synchronous circuit by a set of asynchronous controllers that implement a handshake protocol.

The controllers generate the enable signal for the registers depending on the request/acknowledge signals from successors and predecessors, and a delay matching the combinational logic paths. The problem of clustering memory elements comes from the impossibility to have one controller for each latch of the circuit, and from the sub-optimal solution of having just one controller for all latches.

In this context, the number of skew domains corresponds to the number of controllers required to generate the enable signals of the latches. Finding a good clustering may contribute to improve the performance of the system.

1.2 Previous work

The complexity of assigning arbitrary skews for the memory elements comes from the fact that within-die variations affect the buffers required to implement the skews.

In [5], a solution is proposed to mitigate this problem. The large skews are implemented using different clock domains, with each clock implementing the small skew differences. Their solution is based on a heuristic on top of a satisfiability formulation for the critical core. The authors obtain remarkable results at the expense of a very high computational cost. As an example, their approach took about 20 hours to compute the skews of a design with almost 3000 flip-flops.

Clock latencies can also be calculated to minimize the amount of inserted delay to fix hold violations on the datapath [3] or to minimize the maximum skew [6]. The balance of slacks in the critical paths [7, 8, 9] is another important property, since it reduces the number of critical paths and it makes the design less sensible to process variations. It also gives a unique solution where slacks can be interpreted as a



Figure 1: a) Circuit with a valid useful clock skew assignment for period = 4. b) Register graph with delays annotated. c) Slack graph for the given clock latencies. d) All possible balanced clusters

criticality metric. Latencies generated by distributing slacks will be used in this paper also.

In the context of asynchronous circuits, there have been few efforts to cluster memory elements. Hazari [10] proposes a solution where clusters form a linear pipeline system and each cluster communicates only with two neighbors. The approach uses a variation of a breadth-first search grouping with no control on the number of clusters.

Andrikos [11] noticed the impact of clustering in the performance of an asynchronous circuit. Increasing the number of clusters contributes to improve the performance at the expense of area overhead. However, their approach is based on the topological structure of the netlist and is not guided by the performance.

1.3 Contributions of this work

The goal of this work is similar to the one presented in [5]. Given a circuit and a limited number of skew domains, we want to find the memory elements and the skew assigned to each domain that minimizes the clock period of the circuit.

The main contribution of this work is the low computational cost of the approach that allows to solve the problem in affordable running times. This is crucial in the EDA flows that require a significant effort to reach timing convergence.

To cope with the complexity of the problem, the approach uses a multi-level clustering algorithm [12, 13]. During the coarsening phase, the memory elements are grouped according to a skew affinity metric. During the refinement phase, memory elements can be moved to different domains to improve the quality of the solution.

The obtained results are comparable to the ones presented in [5]. However, the computational cost is drastically reduced. As an example, the skew domains for a design with almost 16,000 flip-flops have been calculated in less than 5 minutes with the approach presented in this paper.

2. OVERVIEW

A synchronous circuit (Fig. 1(a)) is composed of combinational logic gates and registers (A,B,C,D). The performance of the circuit is determined by the cycle time of the clock.

The combinational paths can be represented on a regis-



Figure 2: Possible clock latencies according to the current slack. Grouping registers A and B implies a change on C slack.

ter graph (Figure 1(b)). On a traditional system, the clock signal arrives at the same time for all registers. In this situation the minimum period is the delay of the longest combinational path (Period = 6). The period of this circuit can be improved by applying useful clock skew [1], allowing long paths to steal time from short ones. The optimum performance is now determined by the average delay of each cycle, e.g. $D \to C \to B \to A \to D$ has an average of 3, and the worst cycle is $D \to A \to D$ with an average of 4. The cycle period can be reduced from 6 to 4 by adding latencies to the clock. The skew of two registers is the difference between their clock arrival times.

There are many clock latency assignments that achieve an optimal period of 4, e.g. (A = 1, B = C = 0, D = 3), (A = 1, B = 0, C = 1.5, D = 3) etc. For a given assignment, each edge on the register graph has a margin (slack) value calculated from the delay of the edge, the period, and the clock skew between source and destination registers: slack =period-delay+skew. For the system to be correct, all slacks have to be positive. Fig. 1(c) shows the slacks for the skew solution (A = 1, B = C = 0, D = 3).

The implementation of the clock signal is simplified if registers do not have total freedom for their clock latency. A constrained system where only a few skew groups are defined is easier to implement. Figure 1(d) shows all possible 2-cluster combinations for size-balanced clusters with the resulting period. For example, knowing that the longest path is $A \rightarrow D$, if A and D belong to the same cluster they will have 0 skew and the delay between them will determine the period 6. However, the cluster solution $\{A, B\}$ and $\{C, D\}$ is able to keep the best possible period 4, with the constraint of having just 2 clusters.

This paper presents a fast clustering algorithm that uses skew and slack information to find a clustering solution with good performance. This information can be represented in a time-line diagram. For example, latencies and slacks of Figure 1(c) can be represented as intervals in Figure 2(a). On this scenario, B has a latency of 0, and according to its slack, the latency could go from -1 to 1 without breaking any constraint. Having negative latencies is not a problem because relative distances is what matters. All register latencies can be shifted to be positive. On this example, Aand D are critical and their latency cannot be changed individually, C and D have some slack and the latency can be changed to match another register's latency. Looking at the possible cluster solutions from Figure 1(d), the time-line shows why grouping A and D is not a good idea because the intervals do not overlap, and why $\{A, B\}$ and $\{C, D\}$ seems to be more adequate because they overlap.

Figure 2(b) shows the latencies when A and B are grouped together. Changing the latency of B from 0 to 1 has increased the slack of C.

The presented algorithm uses this information to decide which registers to group together to impact as less as possible the circuit performance. The intervals depend on the initial latencies and any change of a latency affects the neighbor register slacks.

3. PERFORMANCE

This section introduces the necessary theory to evaluate the performance of a circuit. It also defines properties that are used later by the multi-level clustering algorithm.

3.1 Register graph model

A circuit can be modeled as a directed graph $G = (V, E, \delta_{min}, \delta_{max}, P, L)$ where each vertex represents a register, an input, or an output. Each edge represents all combinational paths that exists between two vertices annotating the minimum and the maximum delay of these paths.

- V and E are a set of vertices and edges.
- $\delta_{min}, \delta_{max}: E \to \mathbb{R}^+$ are the minimum and maximum delay of each edge.
- $P \in \mathbb{R}^+$ is the period of circuit.
- $L: V \to \mathbb{R}^+$ is the latency of the clock at each vertex.

3.2 Setup and hold timing constraints

The circuit $(V, E, \delta_{min}, \delta_{max}, P, L)$ is valid if the following *setup* and *hold* constraints are satisfied, and the period P is optimum if it does not exist any lower period and L that satisfy all constraints.

Setup Constraints (no zero-clocking) make sure that all edges $(u, v) \in E$ have enough time to stabilize the result before $v \in V$ stores the result.

$$\forall_{(u,v)\in E} : L(u) + \delta_{max}(u,v) \le P + L(v)$$

Hold Constraints (no double-clocking) make sure that the data coming from $u \in V$ does not overwrite the previous data before $v \in V$ stores it.

$$\forall_{(u,v)\in E} : L(u) + \delta_{min}(u,v) \ge L(v)$$

3.3 Difference constraint graph

A difference constraint multi-graph $G^C = (V, C, \lambda, L)$ can be constructed from the register graph $G = (V, E, \delta_{min}, \delta_{max}, P, L)$ and the setup and hold constraints.

- C is a multi-set of edges $C_s \cup C_h^{-1}$
 - $-C_s = E^T$ is a set of edges generated from the setup constraints (in opposite direction). $\forall (u,v) \in E : \exists (v,u) \in C_s$
 - $-C_h = E$ is a set of edges generated from the hold constraints. $\forall (u, v) \in E : \exists (u, v) \in C_h$
- $\lambda: C \to \mathbb{R}$ represents the maximum difference between arrival times: $L(u) L(v) \leq \lambda(u, v)$. Calculated as:
 - Setup Constraints $L(u) L(v) \le P \delta_{max}(u, v)$ are represented with edges $(v, u) \in C_s$:

 $u \xleftarrow{\lambda(v,u) = P - \delta_{max}(u,v)} n$



Figure 3: Difference constraint graph. Solid and dotted represent hold and setup edges respectively.

- Hold Constraints
$$L(v) - L(u) \leq \delta_{min}(u, v)$$
 are
represented with edges $(u, v) \in C_h$:

 $u \xrightarrow{\lambda(u,v) = \delta_{min}(u,v)} v$

A difference constraint system is feasible if no negative cycle exists on the graph. This problem can be solved using a shortest path algorithm like Bellman-Ford [14]. This algorithm assigns values to the vertex that satisfy the system (if it is consistent) with complexity $O(V \times C)$. These values correspond to the latency L(u) of each vertex $u \in V$.

Figure 3 shows the difference constraint graph created from the example of Figure 1(b). Minimum delays for this example are 1 for all edges, and the λ value for setup edges is specified in terms of the period. On this example, there is a cycle between D and A with a value of P - 6 + 1. This cycle determines a minimum period of 5. Notice how the minimum period without hold constraints was 4.

3.4 Skews and Slacks

3.4.1 Skew

The skew between any two registers is the latency difference of the clock. $K: V \times V \rightarrow \mathbb{R}: K(a,b) = L(b) - L(a)$

3.4.2 Slack

The slack of and edge is the margin for a skew increment without violating any timing constraint. $S: C \to \mathbb{R}$: $S(u, v) = \lambda(u, v) - K(u, v)$

Figure 1(c) shows a circuit graph with slacks annotated on the edges.

For a circuit to be valid, all slacks have to be positive. Increasing the latency of a vertex L(u) by t implies increasing the slack of all successor edges and decreasing it from all predecessor edges by t on the constraint graph.²

3.4.3 Slack interval

The slack interval of a vertex represents the range of latencies the vertex can have without violating any constraint. The interval is the following *interval* : $V \to \mathbb{R} \times \mathbb{R}$:

$$interval(u) = \left(L(u) - \min_{\forall i}(S(u,i)), L(u) + \min_{\forall i}(S(i,u)) \right)$$

meaning that the latency can be shifted until it consumes the minimum slack of the predecessor or successor edges.

3.4.4 Slack Overlap

The overlap of two intervals of two vertices represents the possible latency values than both vertices can be assigned. The overlap is a metric that quantifies the affinity between

¹A constraint graph has multiple edges between two vertices, for setup and hold constraints. Figure 3 shows an example with multiple edges between A and D.

 $^{^{2}}$ Notice that the setup edges on the constraint graph go into the opposite direction.



Figure 4: Circuit graph with distributed slacks. The minimum predecessor slack is equal to the minimum successor slack.

vertices. The overlap $overlap:V\times V\to \mathbb{R}$ can be positive or negative and is defined as:

$$overlap(a, b) = min(r_a, r_b) - max(l_a, l_b)$$

where $interval(a) = (l_a, r_a)$ and $interval(b) = (l_b, r_b)$.

Figure 2 shows an example with different intervals and Figure 5 shows an example with a negative overlap.

3.4.5 Distributed Slacks

A constraint graph is said to have distributed slacks when the minimum predecessor slack is equal to the minimum successor slack for all vertices, i.e.,

$$\forall u \in V : \min_{\forall i}(S(u,i)) = \min_{\forall i}(S(i,u))$$

In this case, all latencies are in the middle of their *interval.* This latency assignment is unique and the slacks and intervals provide a criticality metric for edges and vertices.

Figure 4 shows the distributed slacks solution for the circuit of Figure 1(b). Figure 1(c) represents a non-distributed solution.

3.5 Optimal period

The optimal period of a circuit is the minimum period that satisfies all *setup* and *hold* constraints. There are two interesting results depending on the used constraints.

3.5.1 Only setup constraints

Let us simplify the problem assuming that it is possible to add delays on the minimum paths to fix all hold violations without increasing the maximum delay path [2, 15]. The period with just setup constraints is equivalent to calculate the *Maximum Mean Cycle (MMC)* of the graph using max delays. The most popular algorithms are Karp's algorithm [16], Lawler's algorithm [17], Burns' algorithm [18], and an improved version of Howard's algorithm [19]. A report [20] compares all of them in which Howard's algorithm appears to be the fastest one in practice.

3.5.2 Setup and hold constraints

The problem of finding the minimum period that satisfies all setup and hold constraints can be solved using linear programming [1, 21, 22], or by binary search on the difference constraint graph [23, 24, 6].

This paper uses the binary search approach on the period. A period P is valid if the difference constraint graph $G^C = (V, C, \lambda, L)$ has no negative cycles. The upper bound of the period is the biggest maximum delay (the period without useful clock skew), and the lower bound is the MMC of maximum delays (ignoring hold constraints).

4. CLUSTERING PERFORMANCE

This section studies the implications of clustering on the circuit performance. The period of a cluster solution can be



Figure 5: Slack representation of registers a and b, and their predecessors and successors ap, as, bp, and bs. L' represents the new latency.

calculated using the same graph theory of section 3. A cluster graph can be built by collapsing all registers of each cluster and keeping the maximum and minimum delay between clusters. The complexity of creating this reduced graph is linear.

4.1 Period implications on clustering

Incremental clustering refers to the process of reducing the circuit graph by grouping vertices. These vertices will be part of the same cluster. The idea is to use the *slack* and *overlap* information to calculate the impact on the resulting period without having to recalculate the full graph.

The following theorem gives an upper bound of the period using just the *overlap* information between vertices.

THEOREM 1. Let $G = (V, E, \delta_{min}, \delta_{max}, P, L)$ be a circuit with P being an optimal period for G. Let $a, b \in V$ and $G' = (V, E, \delta_{min}, \delta_{max}, P', L')$ represent the same circuit with L'(a) = L'(b) and P' being an optimal period for G' (no constraints are imposed on L' for the rest of vertices) then,

a) if $overlap(a, b) \ge 0$ then P' = P

b) if
$$overlap(a, b) < 0$$
 then $P \le P' \le P + \frac{-overlap(a, b)}{2}$

PROOF. The proof is divided into two possible situations. Constructing the best case situations:

a) $overlap(a, b) \geq 0$ means that L'(a) has the freedom to match L'(b) without violate any setup constraint. L'(a) does not need to move beyond its interval interval(a). The optimal period is still valid P' = P.

b) overlap(a, b) < 0 means that L'(a) or L'(b) will be moved beyond its *interval*(a) or *interval*(b), generating a negative setup slack. To have a valid system, this negative slack will have to be compensated by increasing the period. We only need to consider the predecessor and successor edges with worst slack. We call these worst adjacent vertices ap, as, bp, and bs (see Fig. 5).

To find an upper bound for the period, assume that all latencies will remain the same except for a and b, $L' - \{a, b\} = L - \{a, b\}$. The best assignment for L'(a) and L'(b) is the middle point of the intervals. Figure 5 shows the initial configuration and how the L' is updated to minimize the impact on the period.

Assuming that both critical edges are from setup constraints (hold constraints are period-independent) the best new latencies for L'(a) and L'(b) are in the middle of the overlap (Fig. 5). The new latencies violate the setup constraints as less as possible.

$$overlap(a, b) = L(a) + S(a, a_p) - L(b) + S(b_s, b)$$
$$L'(a) = L'(b) = \frac{L(a) + S(a, a_p) + L(b) - S(b_s, b)}{2}$$

As all slacks have to be positive for a valid system, the period has to be increased to fix the setup constraints, i.e.

$$S'(a, a_p) \ge 0 \quad \land \quad S'(b_s, b) \ge 0$$

The negative slack is overlap(a, b)/2 for both registers:

$$P' = P - L(a) - S(a, a_p) + \left(\frac{L(a) + S(a, a_p) + L(b) - S(b_s, b)}{2}\right) + S'(a, a_p)$$
$$P' = P + \frac{-overlap(a, b)}{2} + S'(a, a_p)$$

As the slack cannot be negative, the period needs to be increased.

$$P' = P + \frac{-overlap(a,b)}{2}$$

The optimal period is equal to P' for the case were only a and b latencies are changed. This period could be improved by recalculating latencies for all registers. P' represents a upper bound of the new circuit if a and b are grouped and all latencies can be updated. Therefore,

$$P \le P' \le P + \frac{-overlap(a,b)}{2}$$

5. MULTI-LEVEL CLUSTERING

A multi-level clustering approach [12, 13] has been chosen to cluster registers. The technique combines two heuristics to search for the optimal clustering solution. Figure 6 illustrates the technique. During the coarsening phase the graph is reduced by grouping vertices until the size is small enough to easily create the desired clusters. These clusters are propagated back at each level and a second heuristic is used to refine the cluster by moving vertices from one cluster to a better one.



Figure 6: Multi-Level technique

A multi-level technique has shown to be very good for large clustering problems. To avoid accumulating errors on the slacks and the period, they are re-calculated for each level.

5.1 Preliminary

This algorithm can be customized to use *setup* and *hold* constraints or only *setup* constraints. Using hold constraints doubles the number of edges of the constraint graph. If only

Algorithm 1: MultiLevel(G,k)
\mathbf{Pre} : G is a register graph; k is the number of clusters
Post : G is a k- clustered register graph.
begin
if Size(G) < thresholdSize then
DistributeSlacks(G)
list \leftarrow SortVerticesByLatency(G)
CreateSimplePartitions(G,list,k)
else
$GR \longleftarrow Coarsening(G)$
MultiLevel(GR,k)
ProjectClusters(G,GR)
Refine(G)
end

setup constraints are considered, the period can be calculated using Howard's algorithm [19] for the maximum mean cycle (MMC). If hold constraints are also considered, the MMC is calculated as an upper bound for a binary search changing the period on the constraint graph.

Distributing slacks is a very important step on this algorithm, it gives a unique slack assignment where each vertex is assigned at the middle point of its slack interval. The latencies are also assigned in a way that the total number of critical paths (slack=0) is minimized. The slack interval can be interpreted as a global criticality/freedom metric for the vertex. This latency assignment has been used before [8, 9].

A parametric shortest path algorithm [7] can be used to calculate the period and also distribute the slacks. The algorithm decreases the parameter, which is the period and affects only setup constraints, until it appears a negative cycle. At this point, the optimal period has been found. This algorithm can continue by extending the parameter to all edges and collapsing the negative cycles.

5.2 Recursiveness

Algorithm 1 shows the main structure of the recursive function MultiLevel.

When the input graph is small, a simple solution is constructed using latencies after distributing slacks. Grouping vertexes closer in latency will minimize the impact on the period. A simple solution consist of sorting the latencies and divide the list with equally size clusters.

When the input graph is large, the graph is reduced using the coarsening heuristic function on algorithm 2. The reduced graph is clustered by a recursive call and clusters are projected from the reduced graph to the original input graph. Each vertex belongs to the same cluster as his representative in the reduced graph.

The last step of the MultiLevel function is to refine clusters using a second heuristic function. Algorithm 5 explores if some vertices are better assigned into another cluster.

5.3 Coarsening phase

The coarsening phase is a heuristic based on Theorem 1. The goal is to minimize the impact of grouping vertices on the period. Distributed slacks are used because they provide global information for the freedom of the node to change the latency without violating timing constraints. The idea is to keep as much slack as possible for next levels.

Algorithm 2 builds the reduced graph. This is a register graph where each vertex represents a set of FFs and the worst edges are kept between sets of FFs. The algorithm

Algorithm 2: Coarsening(G)	
Pre : G is a register graph; W is the window size Post : GR is a reduced register graph.	
begin	
DistributeSlacks(G)	
$list \longleftarrow SortVerticesByLatency(G)$	
for $i \leftarrow 0$ to Last(list) do	
j ← i	
if list[j] is not already grouped then	
if $i \neq j$ then	
/* if there is a match	*/
NewVertexMerge(GR,list [i], list [j])	
mark list [j] as grouped	
CreateInheritedEdges(GR,G)	
return GR	
ena	

Algorithm 3: FindBestMatch(list, i, W)
$\label{eq:Pre} \mathbf{Pre} \ : \ \text{list is a list of vertices; i is the position of the reference} \\ \text{vertex; W is the exploration window}$
Post : b is the vertex position of the best match or $b = i$ if
there is no match.
begin
b ← i
$best_gain \longleftarrow -\infty$
for $j \leftarrow i$ to $i + W$ do
gain ← CalculateGain(list [i],list [j])
if gain > best_gain and list [j] is not already grouped
then
b ← j
best_gain ← gain
return b
ena

tries to reduce the graph size by half. For each vertex, a best match is searched and merged together in the reduced graph. To avoid the quadratic cost of finding the best match, the search is limited by a constant number W of neighbors. Only vertices closer in latency are visited as candidates for merging.

Algorithm 3 finds the best grouping match for a vertex. It consists of exploring the W closest neighbors and select the vertex with the best gain function. This is an aggressive version of the function. A version were *best_gain* is initialized to 0 gives a more conservative heuristic were only positive gain merges are allowed. This alternative version does not reduce as much the size of the graph and the total number of levels increases, incrementing the computational time.

Algorithm 4 uses Theorem 1 to calculate the implications of merging two vertices. When two vertices are merged, they will be synchronized and they will have the same latency (zero skew). In terms of slack after changing the latency, the resulting graph will have twice the overlap slack (one for each vertex) minus the slack out of the overlap slack (one for each vertex) minus the slack out of the overlap that will be lost for each vertex. This function is a greedy heuristic because it does not take into account the slack implication of adjacent vertices.

Figure 2 shows a situation with intervals and the unpredicted consequences of grouping two vertices.

5.4 Refinement phase

The refinement algorithm 5 tries to improve the period of the cluster solution. The coarsening phase takes inde-

Algorithm 4: CalculateGain(u, v)
\mathbf{Pre} : u and v are vertices to be grouped
Post : gain is the value of a heuristic function that evaluates
the result of grouping u with v.
begin
<pre>/* is the min latency a vertex can have without</pre>
breaking any constraint, r is the max latency.
*/
$overlap \leftarrow Min(u.r,v.r) - Max(u.l,v.l)$
$range \leftarrow Max(u.r,v.r) - Min(u.l,v.l)$
$gain \longleftarrow 2 * overlap - (range - overlap)$
/* The resulting slack will be twice the overlap
minus the cut slack. */
return gain
end

Algorithm 5: Refinement(G)

\mathbf{Pre} : G is a register graph with cluster assignments;
Post: The period of G might be improved by changing cluster
assignments
begin
$CG \leftarrow CreateClusterGraph(G)$
DistributeSlacks(CG)
foreach v in G do
c - ClosestClusterInSlack(CG,v)
if $c \neq v.cluster$ then
AssignCluster(v,c)
UpdateClusterGraph(CG,G)
DistributeSlacks(CG)
end

pendent local decisions that affect the neighbors. The base information for those decisions (the distributed slacks and latencies) is recalculated at the beginning of each level, but the coarsening phase does many merges. Some merges could be a wrong decision depending on the previous merges of the same level.

During refinement phase, the performance of the cluster graph CG is calculated along with distributed slacks. Cluster latencies are assigned to vertices in G and slacks in G are calculated using these new skews. At this point it is possible to detect if a vertex can be reallocated to another cluster in case the slack can be improved by assigning a different latency. After each reallocation the cluster graph CG needs to be updated to recalculate the period and latencies.

Multi-Level architecture makes the refinement phase to reallocate just a few vertices because the right cluster assignment has been driven by successive levels.

5.5 Complexity

For the analysis of complexity, the following variables are considered: v as number of vertices, e as number of edges, and k as number of clusters. The complexity of the algorithm *MultiLevel*(G, k) is determined by the complexity of each level multiplied by the number of levels (log v). The cost of each level is dominated by the sum of costs of *DistributeSlacks* and *Refinement*. The calculation of the optimal period and the distribution of slacks can be done in one run of [7] and it takes $O(ve + v^2 \log v)$. The worst-case complexity of *Refinement* occurs when all vertices are moved to another cluster. This cost can be expressed in terms of the number of clusters k as $O(v(k^3 + k^2 \log k))$, simplified as $O(vk^3)$. In practice, and given the accuracy of the heuristics, the number of moved vertices during the Refinement is small. Therefore, the complexity is typically dominated by the distribution of slacks. The complexity of MultiLevel is $O((v^2 \log v + v(e + k^3)) \log v))$. As the number of clusters is expected to be a small constant, then the cost is $O\left(\left(ve + v^2 \log v\right) \log v\right)$

RESULTS 6.

This section shows how the Multi-Level algorithm performs in terms of quality of the results and execution time. The Multi-Level algorithm provides cluster solutions with a similar quality compared to the current literature. However, the difference with the known algorithms is the execution time for large circuits.

To evaluate the quality of the cluster solutions, ISCAS89 benchmarks have been technology mapped with SIS [25] using the library *lib2.genlib*.

Table 1 shows the comparison between Multi-Level and [5]. The number of vertices is equal to the number of FFs plus two extra vertices, one for inputs and one for outputs. M is the delay of the longest combinational path, which is upper bound for the optimal period. P_{∞} is a lower bound of the period and is the optimum period with as many clusters as vertices. M/P_{∞} gives an idea of the relative period improvement that can be achieved with useful clock skew. Solutions for 2 to 4 clusters have been calculated, showing a relative period compared to the optimum "Cluster Period $/P_{\infty}$ ".

The average values of ISCAS benchmarks show how the period can be improved by using clock skew. The period can be improved compared to the maximum delay M/P_{∞} by 1.14 (14%). For 2, 3, and 4 clusters, the period can be improved to 1.06, 1.02, and 1.02 respectively, showing how with a small number of clusters the period approaches closer to the optimal solution of P_{∞} . The cluster period degradation compared to [5] is about 1%. Notice that some circuits can not be improved like s1196, s510, s641, s713, and s832. Some other circuits can significantly be improved like s382, s400, s444, and s526.

To evaluate the speed of the algorithm several large circuits have been synthesized using commercial tools: three versions of OpenSparc T1 [26], the difference between them is the target period used for synthesis and placement (1, 2, 3ns) and an industrial circuit. The previous 3 largest ISCAS (Table 1) are also included.

Table 2 shows CPU times for these circuits and Table 3 shows the timing results from [5] for industrial large circuits. Figure 7 shows a plot comparing multi-level algorithm versus the SAT based algorithm [5]. Notice that the CPU time is in logarithmic scale.

Tables 2,3, and Figure 7 show the difference in timing complexity. D1, D2, D3, and D8 are taking much more time than the industrial and OpenSparc circuits of Multi-Level. The curve defined by the examples show how Multi-Level runs at two orders of magnitude less time and behaves more scalable for larger circuits.

7. CONCLUSIONS

A new clustering approach for clock skew optimization has been presented. The main goal of the approach is to provide an efficient solution with affordable computational cost.

The combination of multi-level clustering with sound



Figure 7: CPU time vs number of FFs.

heuristics for skew assignment has resulted in a scalable and valuable approach that can be used in large designs. The approach is applicable to either synchronous or asynchronous designs.

In the future, it would be interesting to use the information about the physical location of the sequential elements. Clustering elements that are physically close would contribute to reduce the area and power of the clock trees.

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Circuit						Multi-Level algorithm			Multi-Domain [5]		
						Cluster Period $/P_{\infty}$			Cluster Period $/P_{\infty}$		
name	#V	#E	M	P_{∞}	M/P_{∞}	2cls	3cls	4cls	2cls	3cls	4cls
s1196	19	365	22.28	22.28	1.00	1	1	1	1	1	1
s1238	19	365	26.15	24.33	1.07	1.03	1.03	1.01	1	1	1
s1423	76	2235	79.04	73.13	1.08	1.05	1.04	1.04	1.03	1.01	1
s1488	8	266	23.58	23.18	1.02	1	1	1	1	1	1
s1494	8	266	24.72	23.85	1.04	1	1	1	1	1	1
s298	16	86	13.06	10.79	1.21	1.07	1.07	1.05	1.05	1	1
s344	17	121	15.57	13.14	1.18	1.09	1	1	1.09	1	1
s349	17	121	15.89	13.51	1.18	1.09	1	1	1.09	1.01	1
s35932	1442	6128	289.48	286.32	1.01	1.01	1.01	1.01	1	1	1
s382	23	175	14.07	9.63	1.46	1.20	1.03	1.03	1.20	1.01	1
s38417	1465	31980	87.77	86.19	1.02	1	1	1	1	1	1
s38584	1451	17900	287.78	286.62	1.01	1	1	1	1	1	1
s386	8	129	10.56	9.60	1.10	1.04	1	1	1.10	1	1
s400	23	175	14.6	9.89	1.48	1.18	1.03	1.03	1.17	1.03	1.02
s444	23	175	13.92	8.10	1.72	1.39	1.18	1.15	1.34	1.18	1.10
s510	8	103	14.3	14.29	1.00	1	1	1	1	1	1
s526	23	167	13.48	11.22	1.20	1.07	1.07	1.05	1.05	1	1
s641	21	486	29.99	29.51	1.02	1	1	1	1	1	1
s713	21	486	30.59	30.58	1.00	1	1	1	1	1	1
s832	7	213	16.23	16.22	1.00	1	1	1	1	1	1
average				1.14	1.06	1.02	1.02	1.05	1.01	1.01	

Table 1: ISCAS89 benchmark with setup and hold constraints

 Table 2: Clustering multi-level execution time for large circuits

	μv	μĘ	D	$D_{\rm c}/D_{\rm c}$	CDU(a)
name	# v	#E	Γ_{∞}	Γ_4/Γ_∞	CFU(s)
OSparc 3ns	15849	480324	2.40	1.15	260
OSparc 2ns	15849	480324	1.97	1.01	259
OSparc 1ns	15849	480324	1.45	1.18	240
industrial	7755	267637	3.71	1.12	60
s35932	1442	6128	286.42	1.01	1
s38584	1451	17900	286.62	1.00	1
s38417	1465	31980	86.19	1.00	2

Table 3: Clustering execution time of industrial circuits by [5]

name	#V	#E	P_{∞}	P_4/P_{∞}	CPU(s)
D1	2245	46048	2.79	1.09	12000
D2	2921	250737	15.26	1.04	72000
D3	6316	21006	4.41	1.07	27000
D4	2694	16518	3.69	1.00	1
D5	3065	18030	2.95	1.15	120
D6	574	2294	4.55	1.00	4
D7	852	47370	16.34	1.02	600
D8	2368	9181	1.74	1.05	1800

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