Combining Speculative Execution and Conditional Resource Sharing to Efficiently Schedule Conditional Behaviors

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Scheduling conditional behaviors necessitates the use of a variety of scheduling optimization techniques like conditional resource sharing and speculative execution. Previous research work has clearly shown their effectiveness. The developed heuristics have several drawbacks relating to the effects of syntactic variance on the results. In this paper a list-based scheduling heuristic that exploits conditional resource sharing and speculative execution possibilities, is presented. Its results are quite insensitive to syntactic variance and conditional behavior is effectively accounted for by a probabilistic priority function.

1. Introduction

Scheduling behaviors having a complex conditional structure has been thoroughly investigated in previous research work mainly due to the fact that traditional DFG based heuristics are not adequate to efficiently handle this kind of descriptions [1]. Several better adapted heuristics were proposed [1], [2], [3], [4], [5], [6]. The quality of their results, depends heavily on the ability to exploit conditional resource sharing [2], [4], [6], [7] speculative execution possibilities [3], [5], [16], [17], and shorten path lengths using node duplication techniques [3]. In resource constrained scheduling these techniques permit to better utilize the hardware resources in the datapath and obtain better schedules which result in shorter execution paths and less control logic. An important issue, underlined in previous work [9], [10] relates to the effects of the syntactic variance of the input descriptions, on the synthesis results. These negative effects intervene in two distinct but interrelated levels as far as scheduling conditional behaviors is concerned; mutual exclusiveness detection and scheduling. CDFG based mutual exclusiveness detection techniques (i.e. [3], [11]) using the structure of the input description, produce different schedules for semantically equivalent but syntactically different descriptions due to the variability on the amount of the detected mutual exclusiveness [9]. Furthermore, CFG-based scheduling (i.e. PBS in [6]) is very sensitive to the statement order in the input description. From the above discussion it is clear that in order to efficiently schedule conditional behaviors and possibly optimize the resulting controllers, we have to exploit conditional resource sharing [2], [4], [6], [7] and speculative execution [3], [5], [16], [17], shorten path lengths using node duplication techniques [3] and last but not least cope with syntactic variance [9]. This is the objective of the heuristic described in the following sections.

2. The HCDG representation

The HCDG [20] is a special kind of directed graph that represents data and control dependencies from a data-flow perspective. It consists of the Conditional Dependency Graph (CDG) and the Guard Condition Hierarchy (GCH).

The CDG consists of a set of nodes and set of edges both labeled by guard conditions (called guards in the sequel). Guards are a special type of nodes and represent boolean conditions that guard the execution of operations and the assignment of values to variables. They have also been used as a formal control model in [4], [5] where a CDFG design representation is used. The rest of nodes correspond to operations (i/o, computation, data merging and storage with either register or transparent latch semantics) that compute/assign values to variables. Edges represent data and control dependencies of the nodes.

Guards are equivalence classes of the HCDG elements. For instance, nodes labeled by the same guard are active (carry a value) at the same logical instants. The Guard Condition Hierarchy (GCH) represents the inclusion relations between guards. In [21], it is implemented as a hierarchy of BDD’s. Using BDD’s two things are achieved: first, redundancy is avoided since equivalence between guard formulas can be easily
established, and second, using a factorization process [21] it is easy to find the maximum depth in the tree that a guard can be inserted. Control representations based on BDD’s have already been used in previous work ([15], [4], [5]). The originality of the GCH lies on the hierarchy construction and not at the use of BDD’s which are simply used for their efficiency. Guard inclusion is important in efficiently detecting mutual exclusiveness, independently of syntactic variances. As it was shown in [18], [19] it permits to minimize the number of mutual exclusion tests significantly.

3. HCDG List Scheduling Heuristic

One of the important advantages of list scheduling is that its quality depends on the quality of the priority function [1]. The contribution of this work lies in the choice of the priority function and on the policy of scheduling nodes on the available resources at each iteration of the algorithm.

Scheduling policy outline: At each scheduling step nodes that can be scheduled on the available resources are partitioned into the RDY and SPC, groups corresponding to nodes that can be scheduled without and with speculative execution. Nodes in RDY observe both their data and control dependencies while nodes in SPC fully observe only their data and partially their control dependencies (weak dependencies in [17] is a similar concept). For brevity, the former will be called ready and the latter spec nodes. At each step, ready nodes are scheduled first ([17] has a similar view), exploiting conditional resource sharing possibilities. Once no more ready nodes can be scheduled, resource utilization is increased by further conditionally sharing used resources with spec nodes. Finally, if unused resources still remain we try to utilize them by conditionally sharing spec nodes. Each schedulable node is associated to a schedule guard (sguard) indicating the condition under which the node will execute if it is scheduled in the current control step. For ready nodes the sguard is the guard labeling the node in the HCDG; hence it is statically available. For spec nodes the sguard has to be dynamically computed at each step (similarly to guard smoothing [16], dynamic guards [5], dynamic CV’s [3]) and corresponds to the most profound already scheduled guard ancestor of the spec node’s guard in the guard hierarchy. Using this technique we maximally exploit conditional resource sharing possibilities also for nodes that are speculatively executed.

An adaptive probabilistic priority function: A special priority function was elaborated. For each guard its probability \( p \) is computed from their boolean definition formulas. Using a sum-of-products definition the probability is given as the sum of products of the corresponding boolean variable probabilities. For comparison operations and input booleans a probability of 0.5 is given to both true and false outcomes. A similar technique to obtain predicate probabilities was used in [7]. Next, ASAP and ALAP schedules are used to obtain node mobilities when both control, data dependencies are observed (\( \text{mob}_{c,d} \)) and when only data dependencies are observed (\( \text{mob}_d \)). Node initial weighted urgencies are given by: \( \text{wurg} = p_s \cdot \frac{1}{1 + \text{mob}_d} \). Depending on the case (ready or spec node), either \( \text{mob}_{c,d} \) or \( \text{mob}_d \) is used; \( p_s \) corresponds to the probability of the node scheduling guards (sguards). At the beginning of each scheduling step initial weighted urgencies of ready and spec nodes are adjusted. Adjustment becomes effective once a node has remained unscheduled for a number of iterations higher than its mobility since it became schedulable for the first time. This is achieved by:

\[
\text{awurg}(n) = \text{wurg}(n) + a(n) \cdot p_s(n) \quad \text{where} \quad a(n) \text{ is the adjustment coefficient for node } n:
\]

\[
a(n) = \begin{cases} 
0 \quad \text{if } ((\text{currcstep} - \text{ALAP}(n)) < 0) \\
1 + (\text{currcstep} - \text{ALAP}(n)) \quad \text{otherwise}
\end{cases}
\]

A weighted-projected-resource-demand (wprd) measure was also defined as:

\[
\text{wprd}(n) = \frac{\sum_{t \in R} |R_t| \cdot \sum_{s \in \text{succ}(n,t)} p(h(n,s))}{|R|}
\]

where \( R \) the set of all available resources with cardinality \( |R| \), \( R_t \) the set of available resources of type \( t \) with cardinality \( |R_t| \); \( \text{succ}(n,t) \) are the successors of node \( n \) that will become schedulable for the next control step and can be mapped on a resource of type \( t \); \( h(n,s) \) the guard of the dependency from node \( n \) to node \( s \) and \( p(h(n,s)) \) its probability. The multi-level priority function for a node \( n \) (\( \text{pr}(n) \)) is based on the following sorting keys: \( \text{pr}_1 = \text{awurg} \), \( \text{pr}_2 = p_s \), \( \text{pr}_3 = \text{wprd} \). The first key (\( \text{pr}_1 \)) accounts for the node mobilities in a conditional execution context. The second key (\( \text{pr}_2 \)) further takes into account the conditional nature of the scheduled behaviors giving higher priority to more frequently executing nodes among nodes of equal urgencies. Finally, the third key (\( \text{pr}_3 \)) provides a local measure of how the available resources will be able to accommodate the resource demands provoked by scheduling a node in the current control step.

Conditional Resource Sharing: For each resource type a guard compatibility graph is constructed called
MEG for Mutual Exclusiveness Graph [19]. Each vertex corresponds to a scheduling guard and has an associated list of operation nodes being active under this guard and can be assigned to the resource of that type. Operations in the list are sorted according to their priorities with ready nodes sorted before spec nodes. MEG edges indicate guard mutual exclusiveness and a clique corresponds to a group of pair-wise mutually exclusive guards. A list of nodes that can share a resource is constructed in three steps. First, ready nodes are considered; the guard clique containing the higher priority node and a maximum number of other high priority nodes is iteratively constructed. Finally, for each guard in the clique the first node in it operation list is returned. The best adapted algorithm to find such cliques is based on the initial-graph-partition algorithm presented in [13]. Heuristics like [14] are not as well adapted to satisfy our clique construction objectives since clique maximality is not a good optimization objective in our case.

This process has several advantages. The list-scheduling priority criterion is satisfied for the greatest number of distinct execution instances (paths), simultaneously because the constructed clique contains the highest priority node and the largest number of other higher priority nodes that can share a resource with it. In respect to [9] and [5], speculative execution is considered only after normally executing nodes have been scheduled. In this way the risk of lengthening execution paths by displacing normally executing operations in favor of speculatively executing ones, is avoided. Conditional resource sharing is exploited during scheduling and not before and so lengthening of execution paths due to inappropriate conditional resource sharing (i.e. [2], [11]), is also avoided.

4. Experimental results

The HCDG-based list scheduling heuristic exploiting conditional resource sharing and speculative execution was tested on a set of benchmarks appearing in previous related literature. These are: kim, waka, maha and jian from [2], [7], [12], [8] respectively. For each benchmark the HCDG was constructed, the guard hierarchy was refined, and the HCDG was transformed so that each node is activated only as often as its results are used in succeeding operations. The guard mutual exclusiveness was established using the techniques described in [18]. The HCDG based list scheduling heuristic is compared to other similar heuristics. The obtained results are given in figure 1 for various resource constraints (one cycle resources) in terms of total / longest path / shortest path numbers of states. Published results of other approaches (i.e. Kim [2], CVLS [7], [3], PBS [6], Brewer [5], ADD-FDLS [9]), when available for the particular resource constraints, are also given. Our results are at least as good as the best previously published results. Chaining is applied as a post-scheduling optimization and it is worth noting that without it remarkably good results are obtained.

![Figure 1. Benchmark comparative scheduling results](image-url)
The differences in terms of the higher number of states in shortest paths is explained by the fact that usually the total number of states is lower than in other approaches and so resources are better utilized at each state. Finally, the insensitivity of the scheduling results to the effects of syntactic variance was evaluated (table 1). Our heuristic was applied on two semantically equivalent but syntactically different descriptions (descr.1, descr.2), for each benchmark. The first, has a maximal conditional nesting as opposed to second one where conditions are flattened and each assignment is in its own conditional block. The achieved insensitivity can be attributed first to the dataflow nature of the HCDG (results independent of statement order) and second to the guard hierarchy construction and the mutual exclusiveness detection process that yields the same amount of exclusiveness for both descriptions.

<table>
<thead>
<tr>
<th>Bench.</th>
<th>waka</th>
<th>maha</th>
<th>kim</th>
<th>jian</th>
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<tbody>
<tr>
<td>Resou rces</td>
<td>cmp: 1 cmp: 1 ALU: 2 cmp: 0 cmp: 0 ALU: 2 cmp: 2 cmp: 2 ALU: 2 cmp: 1 cmp: 1 ALU: 1</td>
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Table 1. Insensitivity to syntactic variance

5. Conclusion

Our HCDG-based scheduling approach exploits most of the existing scheduling optimization techniques, enjoying their combined benefits. Both speculative execution and conditional resource sharing are combined in a uniform and consistent framework similar to dynamic CV’s of [3] and guards in [4], [5]. Even more, it does not suffer from effects of syntactic variance at both the mutual exclusiveness detection and scheduling levels, as CDFG, CFG based approaches. Structural, behavioral and data-flow exclusiveness (as classified in [8]) can be extensively detected using graph transformations and guard information. The hierarchical control representation permits to minimize the number of mutual exclusiveness tests and also develop probabilistic priority functions that account for the conditional nature of the design. Experimental results confirm the effectiveness of both the scheduling policy and the proposed priority function.

References