

Efficient Resource Management during Instruction Scheduling for the EPIC Architecture

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Abstract

Effective and efficient modeling and management of hardware resources have always been critical toward generating highly efficient code in optimizing compilers. The instruction templates and dispersal rules of the EPIC architecture add new complexity in managing resource constraints to instruction scheduler. We extended a finite state automaton (FSA) approach to efficiently manage all key resource constraints of an EPIC architecture on-the-fly during instruction scheduling. We have fully integrated the FSA-based resource management into the instruction scheduler in the Open Research Compiler for the EPIC architecture. Our integrated approach shows up to 12% speedup on some SPECint2000 benchmarks and 4.5% speedup on average for all SPECint2000 benchmarks on an Itanium machine when compares to an instruction scheduler with decoupled resource management. In the meantime, the instruction scheduling time of our approach is reduced by 4% on average.

1. Introduction

The Explicitly Parallel Instruction Computing (EPIC) architecture exemplified by the Intel Itanium Processor Family (IPF) relies heavily on compilers to statically schedule instructions to fully utilize its wide execution resources. The majority of instruction schedulers use the dependence critical path lengths as the primary cost function to schedule instructions. Modeling of execution resources is dealt with in an ad hoc way if at all, often in a manner of scattering the hardware details across the entire scheduling phase. The EPIC architecture [11] introduces the notion of instruction bundles and templates, which limit the instruction mixes presented to the hardware so that instructions can be dispatched efficiently to execution units. Instruction templates impose new constraints on instruction packing and dispatching, adding new complexity to resource management. In fact, such new constraints are not unique to EPIC. Modern processors

tend to specialize execution units according to functional types (e.g. memory, ALU, floating-point, etc.) and even have asymmetric functionality within the same types. A given instruction can sometimes be issued only to a particular unit within the same type, and such decisions could even be affected by surrounding instructions.

A simple but sub-optimal approach to deal with these new and complex templates and instruction dispatching constraints is to add a separate instruction-packing phase after instruction scheduling. However, the example in Figure 1 argues for an intelligent scheduling phase equipped with better resource management to achieve optimal performance. Assume a hypothetical implementation of EPIC that can issue three instructions per cycle. It has two memory (M) execution units and two ALU (I) units. For simplicity, assume there are only two instruction templates: MII and MMI, where M and I specify the functional unit types in the respective template slots. The order of slots in a template defines the sequential semantics within a cycle. The string “;” marks a stop bit, the explicit cycle break specified by the compiler to the hardware. Anti-dependences are allowed in the same cycle as long as the instructions are ordered to reflect such dependences.

For the six instructions in Figure 1, a traditional instruction scheduler based on dependence critical paths may derive a two-cycle schedule as shown in Figure 1(a) even if it takes into account the availability of execution units. However, in the decoupled approach, a subsequent instruction-packing phase cannot find any available template to bundle the three instructions in the first schedule cycle, since they require a non-existent MIM template. Reordering the three instructions to use the MMI template is not feasible due to the anti-dependence on y . The bundling phase ends up forcing the “ $ld\ y = [f]$ ” instruction into an extra cycle, resulting in a three-cycle schedule. The bundling phase could attempt to reorder instructions beyond the current cycle with sophistication similar to instruction scheduling. This

ld a = [x]	{mii: ld a = [x]
add b = y, e	add b = y, e
ld y = [f] ;; // can't fit	add x = h, i ;;}
ld c = [g]	{mmi: ld y = [f]
add x = h, I	ld c = [g]
add d = j, k ;;	add d = j, k ;;}
(a) Decoupled scheduling	(b) Integrated scheduling

Figure 1. Decoupled vs. integrated template selection.

would however defeat the purpose of a decoupled approach to separate instruction scheduling from bundling for simplicity. In contrast, if template selection is integrated into the resource management of an instruction scheduler, the optimal scheduling can be achieved using two templates (MII and MMI) in two respective cycles as shown in Figure 1(b).

Instruction scheduling is already one of the most complex phases in an optimizing compiler. Adding in the modeling of instruction templates makes it an even more challenging task. This is further complicated by the instruction dispersal rules that govern the dispatching of an instruction sequence onto execution units at each cycle, the compressed templates for packing instructions from different cycles, the one-to-many mapping possibility from instructions to functional units, etc.

In this work, we propose an extended finite-state automaton (FSA) to model all the resource constraints during instruction scheduling. Each state represents the currently occupied functional units of a cycle and is augmented with instruction templates and dispatching information. The scheduling of an instruction triggers the transition of states. Our extended FSA successfully models the new notion of instruction templates and the set of instructions dispersal rules on the EPIC architecture with both compilation time and space efficiency. Our experimental results show that modeling these additional resource constraints is crucial toward achieving high performance on Itanium. We also encapsulated the modeling and management of hardware resources in a micro-level scheduler to allow easy migration to future implementations of the EPIC architecture or any future processors with similar resource constraints.

In the rest of the paper, Section 2 provides background information and definitions of terminology. Section 3 details the functional-unit based finite-state automaton and its construction. Section 4 discusses instruction scheduling with an integrated modeling of all resource constraints based on the FSA. Section 5 presents the experimental results that compare our integrated approach with decoupled approaches. Section 6 discusses the related work, and Section 7 concludes this paper.

2. Background

2.1 EPIC and terminology

EPIC [11] uses wide instruction words as in the Very Long Instruction Word (VLIW) architecture. There are four *functional unit types* – M (memory), I (integer), F (floating point), and B (branch) on IPF. Each instruction also has an *instruction type* – M, I, A, F, B, and L. The instruction type specifies the functional unit type where an instruction can be executed, where instruction type A (i.e., ALU instructions) can be dispatched to functional units of either type M or I, and instruction type L consumes one I and one F units. Instructions are encoded into bundles where each *bundle* contains three instructions with a specific instruction template. Each instruction occupies one *slot* in a bundle. A *template* specifies the functional unit type for each contained instruction. There are 12 basic templates, such as MII, MMI, MFI, MIB, etc. A *stop bit* dictates that the instructions before and after the stop bit are to be executed at different cycles. Each basic template type has two versions: one with a stop bit after the third slot and one without. Two of the basic templates, MI_I and M_MI, have a stop bit in the middle of a bundle. We call the two *compressed templates* because they allow the packing of instructions from different cycles into smaller code size. On IPF, flow and output register dependences are generally disallowed (with a few exceptions) within a cycle, but register anti-dependences are generally allowed.

Each implementation of the EPIC architecture has its own micro-architectural features. For example, the Itanium processor [12] can issue and execute up to six instructions (two bundles) per cycle. It has a total of 9 *functional units* (FU): 2 M-units (M0 and M1), 2 I-units (I0 and I1), 2 F-units (F0 and F1), and 3 B-units (B0, B1, B2). Functional units under the same type may be asymmetric, requiring certain instructions to be executed only to a particular FU of a FU type. Each processor generation may also have different instruction latencies and execution-pipeline bypasses, resulting in varying latencies between the same pair of dependent instructions when they are dispatched to different FUs. Each processor generation has its own set of *instruction dispersal rules* that describes how each instruction is dispersed to a FU in an instruction sequence. Depending on the bundle location (the first or second in a cycle) and slot location in an instruction fetch cycle, the same instruction may be dispersed to different FUs. How one instruction is ordered or aligned could force another instruction intended for the same cycle to be stalled to a later cycle due to conflict in critical FUs. On Itanium, in a cycle started with an MII bundle, the first I instruction always goes to the I0 unit. If the second I instruction can be executed only on I0, which is already taken, it will be forced into the next cycle. A

detailed description of all of these micro-architectural features for the Itanium processor can be found in [12].

2.2 High- and micro-level instruction scheduling

Our integrated instruction scheduler divides the task of instruction scheduling into two – the high-level instruction scheduling and the micro-level scheduling. The high-level instruction scheduling determines the issue cycle of each instruction according to dependences and instruction execution latencies. It applies ILP (instruction-level parallelism) enabling transformations such as speculation, sets the scheduling priority of instructions, and decides when an instruction is to be executed. The micro-level scheduling takes care of the placement of instructions and resource management within an execution cycle, based on the intra-cycle instruction dependences specified by the high-level scheduler. The micro-level scheduling shields the complicated resource management of the EPIC architecture from high-level scheduling algorithm, making migrating to a different generation of EPIC processors easier.

Both the high-level and micro-level instruction schedulers retrieve machine-specific information from the machine description structures prepared off-line by a machine model builder (MM builder). The machine description structures specify the numbers and types of machine resources (such as machine width, functional units, and templates), the instruction latency, the pipeline bypass constraints, etc. The functional-unit based FSA is also part of machine description structures built off-line. The machine description structures become part of the static data in the compiler executable after the compiler is built. The MM builder constructs the machine description from a published parameter file [13] that describes the micro-architecture details of an Itanium processor.

3. Functional-unit based finite-state automata

Finite state automata have been used in several modern instruction schedulers to model the resource contention in execution pipelines [1, 20]. A well-designed FSA often results in a simple and compact implementation and exhibits better efficiency in both space and time.

While the pipeline resource contention of Itanium is simple to model, the instruction templates and dispersal rules introduce new challenges. The latency between a pair of dependent instructions depends not only on the type of source instruction but also on the functional unit where it is executed. The bundle template and the slot location in a bundle determine the functional unit assigned to execute an instruction. Hence selecting the instruction template and placing instructions into the

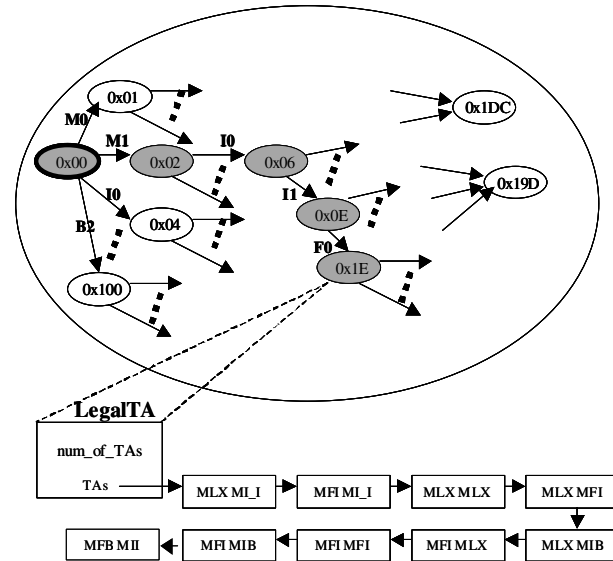


Figure 2. Functional-unit based FSA.

proper slot locations in a bundle is critical toward achieving a high quality schedule.

The importance of template selection intuitively suggests a template-centric model for managing resources. When an instruction is being scheduled, a template-centric model first decides the template assignment for the current schedule cycle and the slot location for the instruction. The instruction latency and the executing functional unit of the instruction are then derived from the instruction template and slot location assigned per instruction dispersal rules. All possible template assignments are enumerated dynamically at each scheduling attempt to select the template and slot.

To improve the compile-time overhead in dynamic template-assignments enumeration, one could build a template-based FSA off-line to guide the selection of template assignments. The template-based FSA would model the template assignment and slot usage of an execution cycle. Each state in a template-based FSA would record all the possible selections of instruction templates under certain slot usage in a single execution cycle. When an instruction was scheduled, an unused slot S would be picked for the instruction and the FSA would transit to the next state where the slot S becomes taken. Template assignment would be selected from the set of templates in the state. Size is one major problem in such a template-based FSA. For the two-bundle wide Itanium, there are at least 68 possible template assignments in a cycle, after accounting for the availability of functional units. The theoretical upper bound on the number of states in a template-based FSA is 2^{68} , the size of the power set of all possible template assignments. Even with aggressive trimming, a template-based FSA still needs dozens of thousand states.

To achieve efficiency in both space and time, we take a functional-unit-centric approach. At the core is a Functional-Unit-based FSA (or FU-FSA in short) that is built off-line. Each state in the FU-FSA represents the set of functional units that are in use (FU usage set) in a schedule cycle. Instructions scheduled into a schedule cycle are treated as a sequence of FU-FSA state transitions. Instruction templates and dispersal rules are modeled in each state by a list of legal template assignments of the state. A template assignment is legal for a state if its FU usage set is a superset of the FU usage set of the state. Only states with at least one legal template assignment are included in the FU-FSA.

When an instruction is being scheduled, the micro-level scheduler looks for a FU P for the instruction that can lead to a legal transition in the FU-FSA. A state transition is legal if the new FU usage set is a valid FU-FSA state and at least one template assignment for the new state satisfies the dependence constraints for all instructions currently scheduled in the cycle per the instruction dispersal rules. When there is no intra-cycle instruction dependence, the legal FU-FSA transition check is simply verifying the new FU usage set is indeed a state in the FU-FSA. The construction of FU-FSA guarantees that at least one legal template assignment exists for each state. When there are intra-cycle dependences in the schedule cycle, the legal FU-FSA transition check needs to ensure the existence of at least one template assignment that can lay out instructions in the required dependence order. This is accomplished by scanning the list of legal template assignments of the new FU-FSA state. In either case, the final selection of template assignment is needed only when scheduling for a cycle has completed, instead of done at every scheduling attempt.

The FU-FSA is much more space efficient than the template-based FSA. The FU-FSA for the nine-FU Itanium has at most 2^9 states, much smaller than that of a template-based FSA. After excluding states with no legal template assignment, the FU-FSA for Itanium contains 235 states. Each state has no more than 38 legal template assignments, while 75% of the states have less than 10 legal template assignments. Therefore the FU-FSA is very compact in term of memory usage and is highly scalable to a wider machine.

Figure 2 illustrates a FU-FSA where the FU usage set in each state is represented as a bit vector. When a FU is assigned to an instruction during instruction scheduling, the current state transits to a new state by setting the bit corresponding to the newly occupied FU. For example, if the current state is 0x00 and the new instruction occupies the M0 unit (bit 0 in the bit vector), the new state will be 0x01. Each state has a structure *LegalTA*, which contains a list of legal template assignments (TAs) for that state per instruction dispersal rules. Figure 2 also highlights a

```

BuildFSA {
  FOREACH FU usage set PV DO {
    FOREACH template assignment T with
      at most 2 bundles DO {
      TV = FU usage set of T;
      IF (PV is a subset of TV) {
        IF (PV is not in FSA) {
          Add PV as a new FSA state;
        }
        Add T to FSA[PV].TAs;
      }
    }
  }
  IF (PV is in FSA) {
    Sort FSA[PV].TAs according to
      priority criteria;
  }
}
Build FSA transition table;
}

```

Figure 3. Building an FU-FSA for Itanium.

possible sequence of transitions of the FU-FSA from the initial state 0x00 to the state 0x1E, following the transition path M1->I0->I1->F0.

When instruction scheduling for a cycle has completed and it is time to finalize the template selection for the cycle, the list of legal template assignments of the current FU-FSA state is scanned to find a template assignment that can best realize the set of instructions in the cycle. One simple approach is to pick the first template assignment in the list that satisfies all required constraints. By properly arranging the list of template assignments for each state during the FU-FSA construction, we can optimize the template selection to minimize code size (or bundle count). A smaller code size in general leads to better performance due to a reduction in I-cache misses. For instance, to minimize code size for Itanium, we sort the list of template assignments according to the following priorities:

1. Ascending bundle count.
2. Putting template assignments with compressed template(s) ahead of full template(s).

Figure 3 outlines the algorithm for constructing the FU-FSA for the two-bundle wide Itanium. It can be easily extended to generate a FU-FSA for any EPIC processor that can issue N bundles per cycle. The algorithm enumerates all possible FU usage patterns, represented as a FU usage set PV . For each PV , we scan all possible template assignments of up to 2 bundles. If a template assignment T has a FU usage set TV per instruction dispersal rules and TV is a superset of PV , T is a legal template assignment for PV . In this case, PV is added to the set of FU-FSA states. The template assignment T is also added to the list of legal template assignments (TAs) for the state PV , that is, $FSA[PV]$. After all template assignments for the FU usage set PV have been examined,

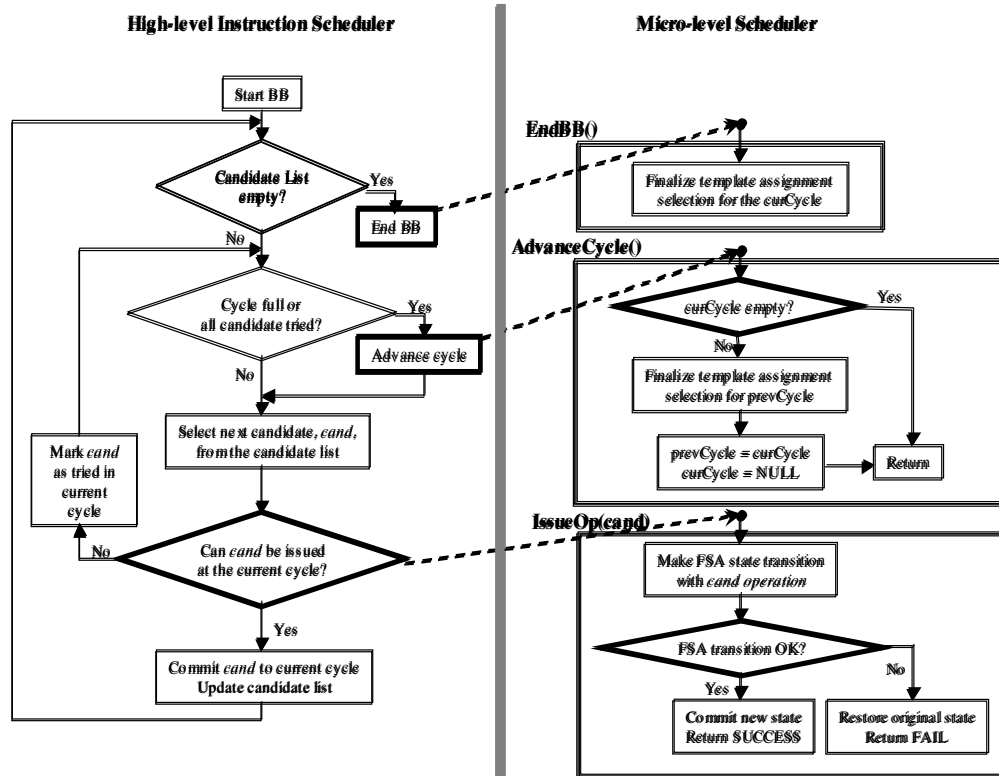


Figure 4. Interaction between high-level and micro-level instruction scheduling.

the list of legal template assignments for *PV* is sorted according to the priority functions described above. Finally the FU-FSA transition table is constructed after all legal states of FU-FSA are included.

4. Instruction scheduler with integrated resource management

With a FU-FSA based micro-level scheduler taking care of the resource management, the instruction scheduler can focus on high-level scheduling decision. Figure 4 highlights the interaction between the high-level instruction scheduler and the micro-level scheduler. On the left-hand side is the flow of a typical instruction scheduler. The right-hand side shows the functions done in the micro-level scheduler.

The instruction scheduler repeatedly picks the best candidate from a list of instructions that are ready for execution at the current cycle. It then consults the micro-level scheduler through the *IssueOp* function to check for resource availability. The *IssueOp* function determines whether there is a FU available for the candidate instruction and whether there exists a legal template assignment that satisfies instruction dependence constraints if exist. Using the FU-FSA, it simply picks an

available FU for the candidate instruction and sees whether the FU results in a legal FU-FSA state transition.

If *IssueOp* completes successfully, the candidate instruction is committed to the cycle. If *IssueOp* is not able to fit the instruction in the current cycle, the instruction scheduler marks the candidate instruction as being tried already. In either case, the scheduler picks another candidate instruction from the candidate list to schedule until the current cycle is full or there is no more candidate instruction to try. The instruction scheduler then closes the current cycle and advances to the next cycle.

The instruction scheduler now needs to decide when to finalize the template assignment for each cycle. One alternative is to finalize the template assignment on-the-fly as soon as scheduling for a cycle is completed (the 1-cycle template selection heuristic). The template selection algorithm looks at the one-cycle window in making template assignment. Or one may defer the template assignment until the whole schedule region is completely scheduled. Selecting the template assignments for several cycles at once allow better packing of instructions in adjacent cycles by exploiting compressed templates. However, it requires extra compilation time and a larger space for maintaining the states of several cycles.

Our instruction scheduler employs a 2-cycle template selection heuristic that is able to utilize compressed templates for better instruction packing while incurs

```

IssueOp(op, cycle) {
  funcUnits= FUs op can be issued to;
  freeUnits = unoccupied FUs in cycle;

  // Try available FUs first.
  candidateUnits = funcUnits & freeUnits;
  FOREACH FU in candidateUnits DO {
    Record op issued to FU in cycle;
    state = getFSASate(cycle);
    IF (state is valid) {
      IF (intra-cycle dependence in cycle) {
        FOREACH ta in FSA[state].TAs DO {
          IF (ChkCycleDep(cycle,ta)==TRUE)
            RETURN SUCCESS;
        }
      } ELSE RETURN SUCCESS;
    }
    Back out op from FU in cycle;
  }

  // Try permuting FU assignments.
  candidateUnits = funcUnits & ~freeUnits;
  FOREACH FU in candidateUnits DO {
    IF (FU is locked) CONTINUE;
    old_op = cycle->op_in_FU(FU);
    Back out old_op from FU in cycle;
    Issue and lock op to FU in cycle;
    IF (IssueOp(old_op, cycle)== TRUE)
      RETURN SUCCESS;
    Back out op from FU in cycle;
    Record old_op issued to FU in cycle;
  }
  RETURN FAIL;
}

```

Figure 5. Pseudo code for IssueOp using FU-FSA.

minimal compilation time and space overhead. Instead of finalizing the template assignment as soon as the scheduling of a cycle is done, the template assignment is selected with a one-cycle delay to give a window of two schedule cycles for template selection. Only the states of two cycles need to be maintained during instruction scheduling, namely, the previous cycle (*prevCycle*) and the current cycle (*curCycle*). When the high-level scheduler advances a cycle, the *AdvanceCycle* function in the micro-level scheduler is invoked to finalize the template assignment for the previous cycle. By looking at both previous and current cycles when selecting the final template assignment of the previous cycle, it allows the utilization of compressed templates. It results in a good balance between the quality of generated code and the space and time efficiency of the instruction scheduler.

Once scheduling for the basic block is completed, the *EndBB* function in the micro-scheduler is called to finalize the template assignment of both the previous and current cycles.

The pseudo code in Figure 5 illustrates a simplified implementation of *IssueOp* using the FU-FSA. Inside the

IssueOp function, unoccupied FUs are first selected for the new instruction *op*, as shown in the first FOREACH loop over the candidate FUs. When a tentative FU is selected for *op*, the cycle is updated to reflect the assignment. The new FU-FSA state is checked to make sure it is a legal state. Furthermore, if there are dependences among instructions within the cycle, the list of legal template assignments (TAs) for the new state must be examined to ensure at least one legal template assignment exists for the required instruction sequence in the cycle. The *ChkCycleDep* function performs the dependence check given a template assignment and the instructions scheduled in the cycle with their FU assignments.

If *op* is not able to use any of the unoccupied FUs, occupied but valid FUs for *op* are tried next. *IssueOp* will re-arrange the FU assignments for instructions already scheduled in the current cycle to exploit the best resource usage. It involves backing out the FU assignment for one or more instructions and re-arranges the mapping of instructions to FUs. The second FOREACH loop over the candidate FUs performs this FU re-mapping. A FU locking mechanism avoids repeating previously tried combinations to ensure termination of the algorithm.

Heuristics that give higher priority to the most constrained instructions can be applied to reduce the search space during FU re-mapping.

4.1 Software pipelining using FU-FSA

The FU-FSA model can be easily integrated into a software-pipelining scheduler as well. The FU-FSA-based micro-level scheduler works on cycle-level resource management. It relies on the high-level scheduler to provide dependence information (both intra- and inter-cycle) among instructions. For illustration purpose, let us assume that the high-level scheduler for software pipelining is a modulo scheduler. The modulo scheduler would model the modulo resource reservation table as *N* consecutive scheduling cycles in the FU-FSA-based micro-scheduler, where *N* is the initiation interval of the pipelined schedule under construction. When querying micro-level scheduler for resource availability at a schedule cycle *C*, the modulo scheduler will send the query to the schedule cycle (*C mod N*) in the micro-level scheduler. The modulo scheduler also provides the micro-level scheduler information on intra-modulo-cycle instruction dependences that account for both loop-independent and loop-carried dependences. The final template assignments for all the *N* schedule cycles will be considered and selected at once after a final schedule is constructed by the modulo scheduler.

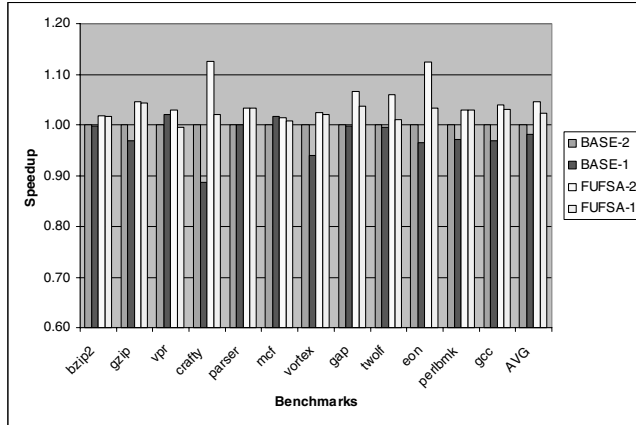


Figure 6. CPU cycles speedup over BASE-2.

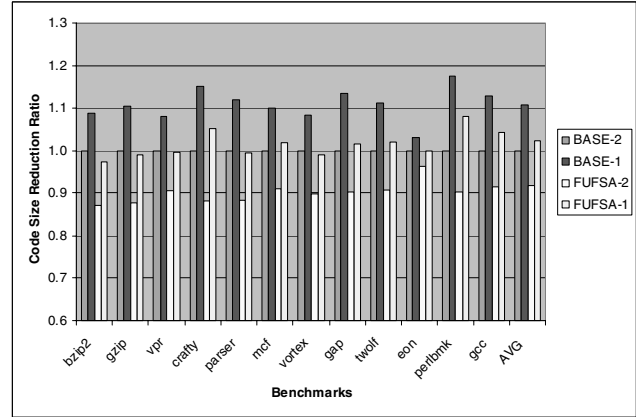


Figure 7. Code Size with respect to BASE-2.

5. Experimental results

The proposed instruction scheduling integrated with FU-FSA-based resource management for the EPIC architecture has been fully implemented in the EPIC Open Research Compiler (ORC) [19]. We compare our integrated approach with an approach of decoupled instruction scheduling and template selection in terms of run-time and compilation-time performance.

ORC includes advanced program optimizations, such as inter-procedural analysis and optimizations, loop-nest transformations, machine-independent optimizations, and profile-guided optimizations and scheduling. The global instruction scheduling reorders instructions across basic blocks. There has been a large amount of research work done on instruction scheduler [2,3,8,9,15,16,17,22]. Our global instruction scheduler uses a forward, cycle scheduling algorithm based on [2]. But it performs on the scope of single-entry-multiple-exit regions, which contain multiple basic blocks with internal control flow transfers. The enhanced cost function is based on the path lengths (to the last cycle) weighted by the execution frequency in a global dependence DAG built for the scheduling region. The global instruction scheduling also utilizes special EPIC architectural features and performs control and data speculation to move load instructions across branches and potential aliasing stores. In case there are spills from register allocation, the local scheduling is invoked for the affected basic blocks. The local scheduling operates on a basic block scope without speculation. Both the global and local instruction scheduling incorporate our FU-FSA based resource management.

We compare two levels of integration in resource management and instruction scheduling for Itanium processors, namely, the decoupled bundling approach (BASE) and our integrated FU-FSA resource modeling approach (FUFSA). Both BASE and FUFSA use the same

instruction scheduler and heuristics. In the BASE configuration, the instruction scheduler performs scheduling based on instruction dependences, instruction latency, pipeline bypass constraints, issue width, and the functional unit availability. But instruction dispersal rules and templates are only taken into account during a subsequent, independent bundling phase that is dedicated to pack instructions under templates. The independent bundling phase invoked after the local scheduling uses the same FU-FSA-based machine model and micro-level scheduler in much the same way as in FUFSA. The bundling phase is not allowed to reorder instructions across the cycle boundary marked by the upstream schedulers.

In the FUFSA configuration, the instruction scheduler incorporates the FU-FSA-based resource management that accounts for instruction latency, pipeline bypass constraints, issue width, types of FUs, templates, and dispersal rules. Instructions are packed under templates on-the-fly when instructions are scheduled without a separate bundling phase.

For each of the BASE and FUFSA configurations, we collected data on two bundling heuristics. The first heuristic (1-cycle template selection) selects the template for instructions in a cycle as soon as scheduling to the cycle is done. It effectively ignores the two compressed templates, MI_I and M_MI. These configurations are called BASE-1 and FUFSA-1 respectively. The second heuristic (2-cycle template selection) defers the template selection of a completed cycle until the next schedule cycle is done, enabling the use of compressed templates to reduce code size. We called these configurations BASE-2 and FUFSA-2. Note FUFSA-2 is the target approach of this work as described in the preceding sections.

We measured performance using all 12 SPECint2000 benchmark programs with full reference input sets. These benchmark programs are compiled using ORC with the peak performance options, which include inter-procedural optimizations, function inlining, profile feedback, and

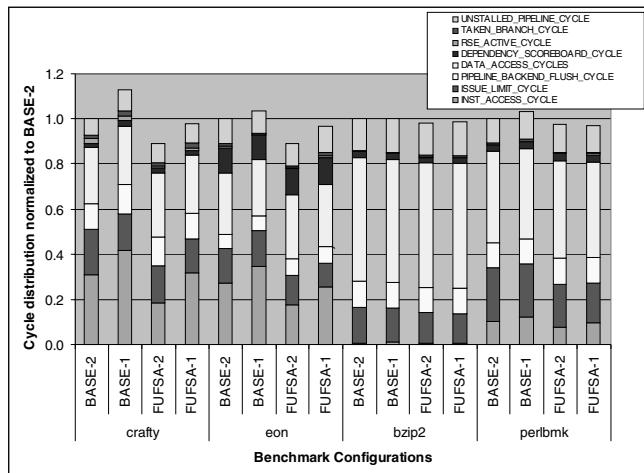


Figure 8. Cycle breakdown for crafty, eon, bzip2 and perlbnk.

extensive intra-procedural and Itanium-specific optimizations. The generated codes are run and measured on a 733 MHz Itanium Workstation with 2 MB L3 cache and 1 GB memory, running RedHat 7.1 version of Linux.

Figure 6 shows the speedup in CPU cycles of all configurations over BASE-2. FUFSA-2 outperforms BASE-2 on all benchmarks, with an average of 4.5% speedup. Crafty and Eon show an impressive speedup of over 12%. We also observed that in general the 2-cycle template selection heuristic performs better than the 1-cycle template selection heuristic. On average, BASE-2 gets a 1.8% speedup over BASE-1 and FUFSA-2 obtains a 2.26% speedup over FUFSA-1.

Figure 7 compares the code size of the four configurations by measuring the size of the text sections in the generated binaries. Note that a shorter bar in Figure 7 indicates a smaller code size. It is clear that FUFSA-2 is able to generate code that is smaller than BASE-2 does. The static code size from FUFSA-2 is reduced by 9.32% with respect to BASE-2. Furthermore we are able to reduce static code size by about 10% when compressed templates are used to pack instructions, as observed from the reduction achieved by BASE-2 over BASE-1 and FUFSA-2 over FUFSA-1.

To further understand how FUFSA improves performance, we use the PFMON (version 0.06) tool to measure dynamic execution statistics through the Itanium performance monitors. Due to space limitation, we select two programs, crafty and eon, which benefit the most from the FUFSA approach, and two other programs, bzip2 and perlbnk, which receive only small improvements from the FUFSA approach. Figure 8 shows the distribution of dynamic cycles under the Itanium stall categories for each of the four programs. Readers are referred to [12] for the full description of each stall category. The cycle distributions for each configuration

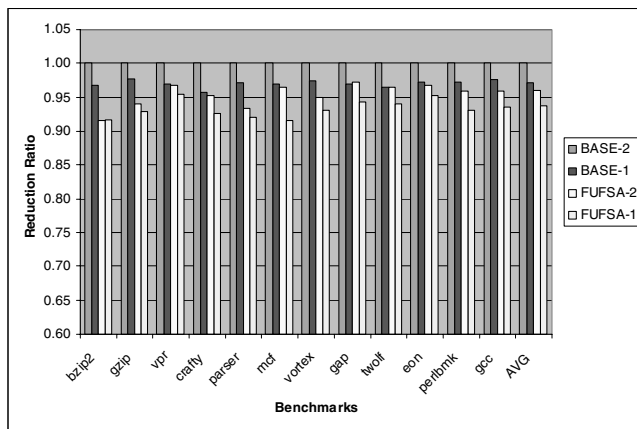


Figure 9. Scheduling time with respect to BASE-2.

have been normalized with respect to the total cycles of BASE-2.

As shown in Figure 8, FUFSA mainly reduces cycles in two stall categories – ISSUE_LIMIT_CYCLE and INST_ACCESS_CYCLE. The ISSUE_LIMIT_CYCLE counts all cycle breaks that are due to the explicit insertion of stop bit in the generated code or the implicit insertion of stop bit by the processor when resources are oversubscribed. The BASE configuration does not account for the constraints of instruction templates and dispersal rules during instruction scheduling. Thus it is more aggressive in scheduling instructions into certain cycles even though there is no instruction template to pack them into a single cycle. The independent bundling phase then needs to split these cycles to fit instruction templates. The FUFSA configuration takes into account the effect of instruction templates and dispersal rules during scheduling, avoiding the template selection deficiency. The FUFSA configurations thus have fewer cycles in the ISSUE_LIMIT_CYCLE categories. On the four benchmarks, FUFSA-2 gets 2-4% speedup over BASE-2 and FUFSA-1 obtains 2-6% speedup over BASE-1 due to the reduction of ISSUE_LIMIT_CYCLE.

FUFSA also shows significant cycle reduction over BASE in the INST_ACCESS_CYCLE stall category, which counts cycles lost to I-cache or ITLB misses. FUFSA-2 gains 12% on crafty and 9.5% on eon over BASE-2 in INST_ACCESS_CYCLE. The reduction in I-cache and ITLB misses can be attributed to the fact that the code generated by FUFSA is more compact than the code from BASE. For benchmarks that spend significant execution cycles waiting for I-cache and ITLB misses, such as crafty and eon, FUFSA is able to achieve higher speedup over BASE by reducing the impact from I-cache and ITLB misses. On the other hand, benchmarks that have fewer cycles in the INST_ACCESS_CYCLE category, such as bzip2 and perlbnk, get a lower speedup from FUFSA over BASE.

We also compare the compilation time spent in instruction scheduling (scheduling time) in Figure 9. We used the cross-build version of ORC hosted on an x86 machine. The compilation time is measured on a workstation with dual 2.4 GHz Pentium IV Xeon processors, 512 KB L2 cache and 512 MB memory. The scheduling time measures the compilation time spent in global scheduling and local scheduling, including the time for the micro-level scheduler and resource modeling. For the BASE configuration, the scheduling time also includes the independent bundling phase for selecting instruction templates. The scheduling time accounts for the majority of the time in the code generator component in the current implementation. The scheduling time at each configuration is normalized to the time at BASE-2. On average, the scheduling time of the FUFSA is about 4% less than the scheduling time of BASE. The scheduling time of using the 1-cycle template selection heuristic is only about 2% less than the scheduling time of using the 2-cycle template selection heuristic. This shows the design of our FSA-based on-the-fly resource management during scheduling provides not only good performance improvements but also compilation time efficiency. It also shows the FUFSA-2 is a better choice in the speedup versus compilation time tradeoff.

6. Related Work

Prior work on modeling hardware resource constraints during scheduling has been mostly on resource (or structural) hazards. Traditional compilers explicitly model the pipeline of the processor by simulating instruction timing. A list of committed resources is maintained in each cycle and tracked by a resource reservation table. Whenever an instruction is considered for scheduling, its resource requirements are checked against the resources already committed in the reservation tables at different cycles. [6, 7, 8, 21] all use the reservation table approach. However, the reservation table approach is less capable of managing instructions that can be handled by multiple types of functional units. Another problem with using resource reservation tables is that the table size is the number of resources times the length of the longest pipeline stage, and every hazard test requires an OR operation on the tables.

Operations in the TriMedia TM1000 mediaprocessor, are issued to different issue slots based their operation types [10]. The mapping from operations to issue slots in TM1000 is fixed. However, the mapping of instructions to FUs in EPIC is determined from the sequence of instructions at each fetch cycle, based on the instruction templates and dispersal rules. The reservation-table-based approach in [4] for assigning issue slots on TM1000 is not capable of handling the template- and context-sensitive resource constraints of EPIC.

FSA-based approach has the intuitive appeal by modeling a set of valid schedules as a language over the instructions. The model in [5, 18] built FSA directly from the reservation tables. The work in [20] reduced the size of FSA by moving away from using reservation vectors as states. Instead each state encodes all potential structural hazards for all instructions in the pipeline as collision matrix. Additional improvements for the FSA-based approach were proposed in [1] to factor, merge, and reverse automata. Note that because the EPIC architecture needs a compiler to model resources primarily at hardware issue time, the FSA in our approach takes advantages of that. Reservation-table-based scheduling usually models resources that are needed at issue time and after the issue cycle.

A recent work [14] is an example of the decoupled approach, and it uses an integer linear programming method to model resource constraints as a post-pass local scheduling on assembly code. A subsequent work [24] extends to model certain aspects of global scheduling though still based on the integer linear programming and post-pass approach. Our scheduling approach may involve backtracking to swap functional unit assignments, which is similar to some modulo scheduling work, e.g. [23]. However, the backtracking in our approach is limited within a cycle, whereas the backtracking in modulo scheduling could go much further.

7. Conclusions

The EPIC architecture and its hardware implementations have introduced a new notion of instruction templates and a set of complicated dispersal rules in addition to the traditional pipeline resource hazards. This has stretched the limit of an optimizing compiler, in particular on instruction scheduling, in its ability to model resource constraints effectively and efficiently in the course of generating highly optimized code. In this work, we have extended the FSA-based approach to manage all of the key resource constraints on-the-fly during instruction scheduling. The FSA is built off-line prior to compilation. To largely cut down the number of states in the FSA, each state models the occupied functional units. State transition is triggered by the incoming scheduling candidate, and resource constraints are carefully integrated into a micro-scheduler.

The proposed scheduling approach integrated with resource management has been fully implemented in the Open Research Compiler. The integrated approach shows a clear performance advantage over decoupled approaches with up to 12% speedup and an average of 4.5% improvement across 12 highly optimized SPECint2000 integer programs running on Itanium machines. This shows the necessity of modeling all resource constraints, including instruction templates and dispersal rules, during

scheduling for a high-performing architecture such as EPIC. We also demonstrate that the compilation time for our integrated approach is competitive to that of a decoupled approach even with a full modeling of the hardware resources. Furthermore, our machine model and micro-level scheduler are modularized and can be easily retargeted to a newer generation of EPIC hardware.

One possible improvement to our implementation is to encode the supported instruction sequences from the legal template assignments in each state. The encoding would allow a faster check on whether intra-cycle dependences among instructions are supported, eliminating the need to walk through the list of legal template assignments. We would also like to investigate incorporating code size as a first-order consideration into our integrated instruction scheduling and resource management model.

Instruction dispersal rules have become ever more complicated on modern processors for both superscalar and VLIW architectures. This is due to various design considerations, such as performance, power consumption, area, and reconfigurability. Our FSA-based approach on scheduling and resource management is a good framework to model such resource constraints during scheduling beyond EPIC. In addition, all VLIW architectures have to pack instructions statically, which can be seen as a form of instruction templates to be modeled by a compiler. We would like to apply our integrated approach to various architectures. One can try to make the set of templates a variable and develop a good schedule for the variable sets of templates, though this remains a challenging problem. We also would like to employ the FU-FSA based approach in a JIT compilation environment where compilation time is critical to the overall performance.

8. References

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