OPTIMIZATION OF EMBEDDED DSP PROGRAMS USING POST-PASS DATA-FLOW ANALYSIS

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ABSTRACT

We investigate the problem of code generation for DSP systems on a chip. Such systems devote a limited quantity of silicon to program ROM, so application software must be maximally dense. Additionally, the software must be written so as to meet various high-performance constraints, which may include hard real-time constraints. Unfortunately, current compiler technology is unable to generate code which may include hard real-time constraints. Consequently, designers often resort to programming application software in assembly. Assembler instructions are highly irregular. Consequently, designers often resort to programming application software in assembly — a time-consuming, error-prone, and non-portable task. Thus, DSP compiler technology must be improved substantially. We describe some optimizations that significantly improve the quality of compiler-generated code. Our optimizations are applied globally and even across procedure calls. Additionally, they are applied to the machine-dependent assembly representation of the source program. Our target architecture is the Texas Instruments' TMS320C25 DSP.

1. INTRODUCTION

It is a well-known fact that the quality of compiled code for embedded DSP systems is extremely unsatisfactory, with respect to density and performance [5]. Consequently, most application software for such systems is hand-written — a very time-consuming, error-prone, and non-portable task. In order to increase developer productivity, DSP compiler technology must be improved substantially.

The overall goal of our research is to develop a retargetable DSP compiler that generates sufficiently dense, high-performance code, hence obviating the need for assembly programming. Our current objective is to identify all optimizations necessary to generate high-quality code for a single architecture — our target is the Texas Instruments’ TMS320C25 DSP [4], a representative of Texas Instruments’ popular fixed-point DSP family. In this paper, we describe some of these optimizations.

Our optimizations are applied globally and inter-procedurally and thus, further improve the quality of locally-optimized code. Additionally, our optimizations occur at the post-pass level — they are applied to the machine-dependent assembly representation of the source program, rather than the machine-independent intermediate representation. It is well-understood that post-pass optimizations which exploit the irregular features of DSP architectures are essential for generating code of the highest quality.

This paper is organized as follows: Section 2 gives an overview of the TMS320C25 architecture. Section 3 describes our optimizations. Section 4 provides experimental results, and finally, Section 5 presents our conclusions.

2. TMS320C25 ARCHITECTURAL OVERVIEW

We first describe the organization of the TMS320C25 memory subsystem, since it motivates the need for global and inter-procedural post-pass optimizations. The TMS320C25 does not penalize the use of the absolute addressing mode. Consequently, for each procedure that is not involved in recursion, the compiler can statically allocate all automatic variables (those local variables that become active on procedure entry and inactive on procedure exit) and refer to them by their absolute memory address. This eliminates instructions to set up stack frames and modify the frame pointer. In contrast, architectures such as the Motorola 56000 incur a penalty for each use of the absolute addressing mode — an extra instruction word is required to hold the absolute address. In these architectures, it is advantageous to allocate the automatic variables on the stack frame and access them indirectly through one or more address registers.

Main memory is divided into 311 pages, with each page subdivided into 128 16-bit words. A 2-bit register DP is dedicated to hold a single page number. When the value of DP is incremented with a 7-bit offset that occurs in the instruction word, a 16-bit address is formed which represents the absolute address of a word in memory. However, before accessing main memory, one must ensure that DP contains the correct page number. This is the purpose of the LDPK instruction, which loads DP with a 9-bit constant.

Consider the instruction sequence of Figure 1 which adds the contents of statically-allocated variable var to the accumulator. Assume that var has been allocated to memory page two at offset 100. The first instruction loads DP with the number of the page containing var. The second instruction first forms the absolute address of var by concatenating the value of DP with the offset of var within page two. It then adds the contents of this memory location to the accumulator. A basic block is defined to be a code sequence in which control-flow enters at the beginning and leaves only at the end [1]. Observe that if two successive memory references in a basic block access variables allocated to the same page and no procedure call occurs between these
references, an LDPK instruction before the second reference becomes redundant. Thus, our objective is to take advantage of the static allocation capabilities offered by the TMS320C55. However, our efforts will be successful only if the LDPK overhead resulting from static allocation is significantly less than the instruction overhead resulting from stack-based allocation. We will now describe how post-pass optimizations may be used to reduce the LDPK overhead.

3. POST-PASS OPTIMIZATIONS

3.1. Reducing the LDPK Overhead with Post-pass Data-flow Analysis

Assembly code is traditionally generated on a per-basic block basis. Lin [6] provides a simple post-pass algorithm that removes redundant LDPK instructions within basic blocks, assuming that an LDPK is generated for each access of a statically-allocated variable. Alternatively, it is possible to suppress generation of many redundant LDPKs during code generation itself. Prior to code generation, it is assumed that the value of DP is unknown. Before code is generated, to access a statically-allocated variable $v$, the value of DP is compared to the number of the page to which $v$ has been allocated. If these values are identical, then generation of an LDPK instruction is suppressed, otherwise the value of DP is updated and an LDPK is generated. It is also assumed that a procedure call destroys the value of DP. Hence, for each basic block, an LDPK is generated for the very first access of a statically-allocated variable and for each access that is the first to follow a procedure call.

By performing global post-pass data-flow analysis [1] for each procedure $p$, we can determine the value(s) that DP assumes upon entry to each basic block of $p$. We may then remove the very first LDPK instruction of each basic block $B$ (assuming it is not preceded by a procedure call) if and only if the value of DP upon entry to $B$ is equal to the number of the corresponding page. In order to do this, we must compute three values for each basic block $B$:

- IN($B$): value of DP upon entry to $B$
- OUT($B$): value of DP upon exit from $B$
- LAST($B$): number of page last accessed in $B$

LAST($B$) is computed by examining the assembly instructions of $B$ in reverse order. We still assume that a procedure call destroys the value of DP, so if a call is encountered before an LDPK in our traversal, LAST($B$) is assigned the constant UNKNOWN. If no LDPKs or procedure calls are encountered, LAST($B$) is assigned the constant PROPAGATE. Otherwise, it is assigned an integer corresponding to the page number referenced by the LDPK. We then compute IN($B$) and OUT($B$) using the equations of Figure 2 - in these equations, a null intersection is equivalent to a value of UNKNOWN. Once a solution to these equations has been found, we may attempt to remove the very first LDPK instruction of each basic block as previously described. Note that due to the global nature of these equations, IN($H$) is assigned the value UNKNOWN, where $H$ is the entry basic block of the current procedure. This implies that the very first access in each procedure of a statically-allocated variable will be preceded by an LDPK. Additionally, an LDPK will still precede each access that is the first to follow a procedure call.

By performing inter-procedural post-pass data-flow analysis, we can determine the value(s) that DP assumes after each procedure call, as well as upon entry to each procedure. This information allows us to potentially remove a subset of the remaining LDPK instructions. The former can be derived by assigning DP the value of OUT($T$) after each procedure call, where $T$ is the exit basic block of the called procedure. The latter can be computed, for each procedure $p$, by examining the value of DP prior to each procedure call to $p$ in the program, and then taking the intersection of these values. Note that DP is assumed to be unknown before program execution begins. Thus, the very first access in the program of a statically-allocated variable will always be preceded by an LDPK.

```plaintext
while (OUT(B)'s have changed) {  
  for each basic block B {  
    IN(B) = \bigcap_{p in predecessor(B)} OUT(p);  
    if (LAST(B) == PROPAGATE)  
      OUT(B) = IN(B);  
    else OUT(B) = LAST(B);  
  }
}
```

Figure 2. Global data-flow analysis equations used for LDPK and RSXM/SSXM removal.

3.2. Reducing the Data-Memory Overhead by Coloring Automatic Variables

The primary disadvantage of statically allocating automatic variables is that the run-time data-memory requirements of the program increase considerably – statically-allocated...
variables are assumed to exist for the entire duration of the program and hence, may not be dynamically deallocated. Reducing the data-memory overhead of the application software is essential in the design of embedded systems.

By definition, automatic variables are required to exist during the time period between procedure entry and procedure exit. This allows us to perform a graph coloring of all statically-allocated automatic variables (henceforth referred to as eligible variables), such that identically-colored eligible variables are allocated to the same data-memory location. In the first step of the coloring process, live-variable analysis [1] is performed on each procedure’s eligible variables so as to determine their live ranges. The live range of a variable $v$ specifies those regions of the program where $v$ is active. We enforce liveness relationships among eligible variables of different procedures using the following heuristic: at each call instruction $s$ in procedure $p$, the live ranges of all eligible variables of $p$ active at $s$ are assumed to overlap with the live ranges of all eligible variables of $(i)$ the called procedure $c$ and $(ii)$ those procedures directly and transitively called by $c$. In the second step, an inter-procedural interference graph is constructed in which a vertex exists for each eligible variable in the program, and edge $e(i, j)$ exists if and only if the live ranges of variables $i$ and $j$ overlap. This edge specifies that $i$ and $j$ must be colored differently, or analogously, they must be allocated to different data-memory locations. Otherwise, a definition of $i$ will overwrite the value of $j$, or vice versa, hence leading to incorrect code.

We now use a binary search to determine a small number of colors that satisfies the graph-coloring constraints imposed by this interference graph (call it $G$). At each step of the search, we apply Briggs’ coloring heuristic [3] to $G$. Experimental results demonstrate that this optimization significantly reduces the total quantity of data-memory consumed by the program. Additionally, if the LDPK optimization is applied after coloring, we find that we are able to further reduce the LDPK overhead — since fewer pages of memory are consumed, DP now assumes fewer values.

3.3. Reducing the RSXM/SSXM Overhead with Post-pass Data-flow Analysis

The post-pass data-flow analysis techniques of Section 3.1 may be used to eliminate other redundant assembly instructions. The TMS320C325 permits the programmer to set and reset the sign-extension (s-e) mode using the SSXM and RSXM instructions, respectively. The former instruction causes future accumulator loads to be automatically sign-extended, while the latter suppresses sign-extension. Let us redefine $IN(B)$, $OUT(B)$, and $LAST(B)$ for each block $B$:

- $IN(B)$: value of s-e mode upon entry to $B$
- $OUT(B)$: value of s-e mode upon exit from $B$
- $LAST(B)$: last value of s-e mode in $B$

$LAST(B)$ is again computed by examining the assembly instructions of $B$ in reverse order: if a procedure call is encountered before an RSXM/SSXM instruction in our traversal, $LAST(B)$ is assigned the value UNKNOWN. If no RSXM/SSXM instructions or procedure calls are encountered in the traversal, $LAST(B)$ is assigned the constant PROPAGATE. Otherwise, it is assigned the constant RESET or SET depending on whether an RSXM or SSXM, respectively, was encountered first.

Using the equations of Figure 2, we now compute $IN(B)$ and $OUT(B)$ for each basic block $B$. Assume that the first sign-extension-mode-related instruction in block $B$ is not preceded by a procedure call. Then if $i$ is equivalent to an RSXM (SSXM) instruction, we may safely remove it if and only if $IN(B)$ has a value of RESET (SET). Inter-procedural post-pass data-flow analysis may be used to remove other redundant RSXM/SSXM instructions.

3.4. Post-pass Optimizations in a Retargetable Environment

The routines responsible for reducing the LDPK and RSXM/SSXM overhead are virtually identical — the only difference occurs in the routines responsible for computing LAST. Rather than implementing these optimizations with two sets of nearly-identical code, we have instead developed generic post-pass data-flow analysis interfaces which these and other optimizations may use. In particular, we have developed generic interfaces for post-pass reaching definitions analysis, live-variable analysis, and available expressions analysis [1]. Consequently, these optimizations have been implemented with very little code duplication.

In order to make use of these data-flow analysis routines, one must provide two optimization-specific functions to the appropriate interface. Each of these functions takes an assembly instruction as input and outputs a bit-vector. For reaching definitions and available expressions analysis, these functions are $gen$ and $kill = gen(i) \& kill(i)$ represent the set of definitions generated and killed, respectively, by instruction $i$. For live-variable analysis, these functions are $use = gen(i)$ and $def = kill(i)$ represent the set of data used and defined, respectively, by instruction $i$. We have used the available expressions interface to implement the optimizations of Sections 3.1 and 3.3. As an example of the use of the $gen$ and $kill$ functions, assume variable $v$ has been statically allocated to data-memory page 5. Then,

- $gen( LDPK(v) ) = \{ 5 \}$
- $kill( LDPK(v) ) = \{ 0, 1, 2, 3, 4, 6, 7, ..., 511 \}$

We have also developed a generic graph coloring interface that performs live-variable analysis and graph coloring of optimization-specific data. We have implemented the optimization of Section 3.2 by providing this interface with the set of eligible variables of each procedure in the application program. We have also used this interface to allocate memory to the pointer variables of a program.

4. EXPERIMENTAL RESULTS

We have implemented our optimizations in the SPA M compiler [5] — a joint project of Synopsys, Inc., Princeton University, University of Aachen, and M.I.T. — which is a retargetable code generation framework for embedded DSP processors. Table 1 shows the results of applying our post-pass optimizations to benchmarks from the DSPStone benchmark suite [7] — a lower level speech-encoding algorithm, while complex update, fft, fir, irix bisqaud, lms and matrix multiply are small DSP kernels. The benchmarks are listed in the first column of Table 1. The next column specifies the
total number of LDPK instructions present in the assembly code after basic block code generation has been performed. These numbers serve as base measurements against which we will compare the results of our optimizations. The next two columns specify the total number of LDPKs present after global and inter-procedural data-flow analysis, respectively, have been performed. The numbers in parentheses specify the total number of LDPK instructions present after variable coloring and post-pass data-flow analysis have been performed in succession. The next column specifies the total number of RSXM/SSXM instructions present in the assembly code after basic block code generation has been performed. Again, these values serve as base measurements. The final two columns specify the total number of RSXM/SSXM instructions present after global and inter-procedural data-flow analysis, respectively, have been performed.

In our experiments, we assumed the compiler allocated eligible variables to data-memory in order of decreasing static frequency. It is evident that inter-procedural post-pass data-flow analysis, combined with coloring of automatic variables, essentially eliminated the LDPK overhead in each of these benchmarks. For instance, for the adpcm benchmark, global data-flow analysis combined with coloring resulted in an LDPK instruction count of 37, a 94.6% reduction from the base measurement of 699, while inter-procedural analysis combined with coloring resulted in an LDPK overhead of just one instruction. These results can be attributed to the fact that after coloring, only one page of memory was required for static storage of eligible variables. For the smaller benchmarks, inter-procedural data-flow analysis resulted in the same LDPK overhead, regardless of whether or not coloring was performed. For example, for the fft benchmark, global data-flow analysis combined with coloring resulted in an LDPK instruction count of 17, a 69% reduction from the base measurement of 55. Inter-procedural analysis with or without coloring reduced the LDPK count to just one. These results can be attributed to the small number of eligible variables present in these programs – the compiler was able to allocate them to one page of memory even without coloring.

Finally, it is evident that our optimizations were quite successful at reducing the RSXM/SSXM overhead. For instance, for the adpcm benchmark, global data-flow analysis resulted in an instruction count of 59, a 43.8% reduction from the base measurement of 105. Inter-procedural analysis reduced the overhead further by 16.9%. For the fft benchmark, our optimizations reduced the overhead from the base measurement of 25 to 5, an 88% reduction.

<table>
<thead>
<tr>
<th>DSPstone Benchmark</th>
<th>LDPK Count (After Coloring)</th>
<th>RSXM/SSXM Count</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Initial</td>
<td>Global</td>
</tr>
<tr>
<td>adpcm</td>
<td>622 (461)</td>
<td>385 (172)</td>
</tr>
<tr>
<td>complex update</td>
<td>4 (4)</td>
<td>4 (4)</td>
</tr>
<tr>
<td>fft</td>
<td>55 (56)</td>
<td>17 (17)</td>
</tr>
<tr>
<td>sir biquad</td>
<td>40 (40)</td>
<td>5 (5)</td>
</tr>
<tr>
<td>ims</td>
<td>13 (13)</td>
<td>6 (6)</td>
</tr>
<tr>
<td>matrix multiply</td>
<td>19 (19)</td>
<td>4 (4)</td>
</tr>
</tbody>
</table>

Table 1. Results of post-pass optimizations.

We have described some post-pass optimizations that significantly improve the quality of compiled code for embedded DSP systems. In particular, we have shown how redundant LDPK and RSXM/SSXM instructions can be eliminated by performing global and inter-procedural data-flow analysis on the assembly code. We have also shown how coloring of all statically-allocated automatic variables can significantly reduce the data-memory requirements of the application software, as well as further reduce the LDPK overhead. Finally, we have described our implementation of generic interfaces for post-pass data-flow analysis and graph coloring which can be used to implement these, as well as other machine-dependent optimizations.

6. ACKNOWLEDGEMENTS

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REFERENCES


