Code Generation and Optimization for Java-to-C Compilers*

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Abstract. Currently the Java programming language is popularly used in Internet-based systems, mobile and ubiquitous devices because of its portability and programability. However, inherently its performance is sometimes very limited due to interpretation overhead of class files by Java Virtual Machines (JVMs). In this paper, as one of the solutions to resolve the performance limitation, we present code generation and optimization techniques for a Java-to-C translator. Our compiler framework translates Java bytecode into C codes with preserving Java's programming semantics, such as inheritance, method overloading, virtual method invocation, garbage collection, and so on. Moreover, our compiler translates for in Java into for in C instead of test and jump for better performance. Our runtime library fully supports Connected Limited Device Configuration (CLDC) 1.0 API's.

1 Introduction

Java's platform independent architecture gives excellent portability. The applications written in Java can be compiled into location-independent codes moving on the Internet and running on every platform. In addition to the portability, its enhanced programability for more advanced development is one of the most important merits. However, despite of the distinguished advantages over other programming languages, there are two shortcomings to use Java, i.e, the size of Java virtual machines and performance limitation due to interpretation. The size of a full-featured JVM is too big to be used on small devices like mobile phones and PDAs. Due to the limited computing power resources on small embedded devices, a few different versions of JVMs have been proposed [1,2], and therefore, the class files cannot be executed on all kinds of client machines. Examples of the full-featured JVMs are JVM from Sun Microsystems [3], Jikes RVM [4], and an example of the partially-featured JVM due to the resource constraint is Java 2 Platform, Micro Edition (J2ME) [2].

The software interpretation incurs much higher runtime overhead than direct execution to use native codes. To alleviate the performance problem, many methods have been proposed such as just-in-time (JIT) and ahead-of-time (AOT)

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compilers. The just-in-time compiler [4,5,6] converts a sequence of bytecode (a method) into native codes at runtime, dynamically links, and executes them. This approach has been widely used for the last years. The alternative method is to translate class files into C codes on offline [7,8,9].

In this paper we present code generation and optimization techniques for a Java-to-C compiler. Also our compiler fully supports Connected Limited Device Configuration (CLDC) 1.0 API's. Moreover, our Java-to-C compiler translates for in Java into for in C in order to get better performance because gcc compiler cannot optimize goto statement generated for for in Java by our earlier Java-to-C compiler, but it can apply various optimization techniques to for in C.

Toba [8] is a system to generate standalone Java applications which were targeted for JDK 1.1. It has a bytecode-to-C translator and additional runtime libraries to support garbage collection, thread management and Java API. In [9], the Java-through-C compilation system for embedded systems has been developed. There is a definite difference between our Java-to-C compiler and the others. Our compiler only generates good quality of codes, to which a backend compiler (ex. gcc compiler) can effectively apply code optimization for improving performance.

The paper is organized as follows: we briefly present the structure of our Javato-C compiler in Section 2, and the framework for C code emission from Java bytecode in Section 3. Also, In Section 4, we present the implementation of one optimization technique which generates for in C for for in Java instead of test and jump. Section 5 discusses issues about code generation and optimization with performance analysis, and finally Section 6 makes the conclusion.

2 Structure of Java-to-C Compiler

Our Java-to-C compiler is organized into three components: a Java decoder, a bytecode-to-C translator, and runtime libraries. The Java decoder analyzes the class files, and generates class blocks to maintain class information. The translator converts Java bytecode sequences into a sequence of C codes. Finally, the generated C codes are linked with runtime libraries to include routines for garbage collection, thread management, and Java CLDC 1.0 API in order to build executable codes. A thorough description of the structure is provided by our technical report [10].

3 Code Generation

There are a few issues in code generation, such as bytecode translation, exception handling, garbage collection, and thread management.

During translation, the preprocessor splits the whole bytecode sequence into several basic blocks to construct a control flow graph (CFG) for a method. The control flow graph is used to compute the stack state. The liveness of temporary variables are properly maintained through computing the stack state. Because bytecode translation is performed in compilation time, the information of the

simulated operand stack is useful for code generation. After the preprocessor finds out basic blocks, bytecode-to-C translation is preformed for each basic block. The generated C codes for all basic blocks in a Java method are wrapped up in a switch statement like Toba [8].

The switch statement is used to handle Java exceptions. Some exceptions are specified to be thrown in Java virtual machine specification [11] when certain conditions are satisfied. These exceptions can be ignored according to the execution environment. The runtime program counter of the JVM is used to find a corresponding exception handler. A local variable pc is employed to mimic the program counter in the JVM. The variable is set at the beginning of every basic blocks. Additionally, C's setjmp and longjmp routines are used to handle exceptions.

An automatic garbage collection is supplied in Java. If certain objects are no longer referenced in a Java program, the objects will be de-allocated without any effort by a programmer. We use Boehm-Demers-Weiser conservative garbage collector [12].

Our system has been targeted to support the Connected Limited Device Configuration (CLDC) 1.0 [13]. Because the only java.lang.Thread class is specified in CLDC API, complex runtime libraries for java.lang.ThreadGroup are not implemented. The Java runtime package for thread management is implemented using Linux PThread libraries. A thread can hold only a lock associated with a monitor which all objects have competed. Synchronization between threads is guaranteed via monitors.

4 Optimization

4.1 Motivation

Figure 1 shows one of major Java methods in LU benchmark, and Figure 2 shows C output generated by our Java-to-C compiler and assembly output by gcc compiler. When the Java-to-C compiler translates Java for loop in forms of test and jump statements instead of for in C, gcc compiler sometimes does not assign a loop induction variable to a register. As shown in Figure 2, the loop induction variable Lvi14 is incremented by leal 1(%eax), %eax instruction. It incurs huge processor stalls. When the Java-to-C compiler generates for in C for for in Java, we could get about 20% higher speedup than before, since gcc compiler applies loop optimizations to for in C. Therefore, we designed a state machine to identify for in Java bytecodes for generating for in C. Using gathered information, our compiler can translate for in Java in forms of for in C instead of test and jump.

4.2 Detection of for Loop in Java Bytecode

Our compiler detects Java for loops by using a state machine to recognize the following code sections in bytecode sequences: an initialization of an induction variable, a loop bound, a modification of an induction variable, and a backward

```
public static int factor(double ad[][], int ai[])
{
    ....
    for (int i2 = 1 + 1; i2 < i; i2++)
        ad2[i2] -= d3 * ad3[i2];
    ....
}</pre>
```

Fig. 1. Example of Java code in LU

```
_L229:
                                            .L184:
    pc = 229;
                                                cmpl
                                                         -204(%ebp), %eax
    i0 = Lvi14;
                                                         $229, -188(%ebp)
                                                movl
    i1 = Lvi2;
                                                         .L186
                                                jge
    if( i0 >= i1 ) goto _L257;
                                            .L187:
                                                         %st(0)
L235:
                                                fld
    pc = 235;
                                                fmull
                                                         16(%esi,%eax,8)
    a0 = Lva10;
                                                         16(\%edx,\%eax,8)
                                                fsubrl
    i1 = Lvi14;
                                                         16(\%edx,\%eax,8)
                                                fstpl
    a2 = a0;
                                                leal
                                                         1(%eax), %eax
    i3 = i1;
                                                         $235, -188(%ebp)
                                                movl
    d2 = ((struct darray*)a2)->data[i3];
                                                         .L184
                                                jmp
    d3 = Lvd12;
                                            .L186:
    a4 = Lva11;
    i5 = Lvi14;
    d4 = ((struct darray*)a4)->data[i5];
    d3 = d3 * d4;
    d2 = d2 - d3;
    ((struct darray*)a0)->data[i1] = d2;
    Lvi14 += 1;
    goto _L229;
_L257:
```

Fig. 2. C and assembly outputs from the example Java code in Figure 1

jump instruction which is the end of for loop. The state diagram is shown in Figure 3. This state machine can recognize several for loop patterns in Java bytecodes, and the recognizable patterns are shown in Table 1.

Initialization of an induction variable: In the state machine of Figure 3, state q0 is an initial state and state accept is a final state. In Java bytecode sequences, an induction variable can be either 1) a local variable or 2) a member field or a method of an object. In Case 1, bytecodes such as iconst, bipush, sipush, load, invokestatic, getstatic and so on, are used to initialize an induction variable. In Case 2, aload appears first, and then arraylength, getfield, invokespecial, invokeinterface, or invokeivirtual follows. If a compiler detects either case, the state machine goes to state q1. When the state machine is in state q1 and the following instruction is istore, the state machine

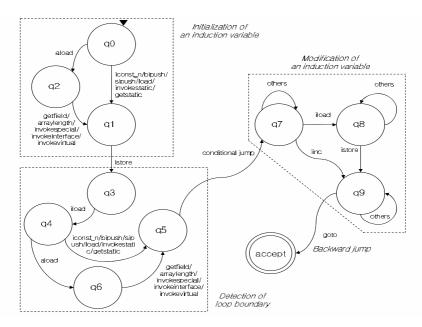


Fig. 3. State diagram to search for in Java bytecode

moves into state q3. This istore instruction stores the initialized induction variable into a memory. Thus, when the state machine reaches state q3, it decides that a code section of an induction variable initialization is found.

Loop bounds: In Java bytecode, after initialization of an induction variable, a code section of a comparison part should appear according to a syntax of for in Java. If each condition for the syntax is not satisfied, the state machine is moved into the initial state q0. To compare an induction variable with a loop bound, iload instruction is performed to read an induction variable from a memory. At this time, the state machine moves from state q3 into state q4. After the induction variable is loaded from a memory, a loop bound variable is loaded. If the loop bound variable is a local variable, the state is moved to q5 from q4, and if the loop bound variable is a member field or a method of an object, the state is moved to q6 and q5 from q4.

Modification of an induction variable: We define two patterns as modification of an induction variable in Table 1 for simple experiment. In the first pattern, we should find a pair of iload and istore instructions that access the same memory location before goto statement for a backward jump ($q7 \rightarrow q8 \rightarrow q9$). The kinds of integer arithmetic and logic operations between these two instructions are not important. All we need are the start and the end of a code section of modification of an induction variable. In the second pattern, the start and the end of a code section of modification of an induction variable are the same. When iinc instruction appears, the state machine moves from state q7 into state q9. The only condition that iinc is an instruction for a code section

$for(A = B; A < C; A = A+D \text{ or } A++) \{\cdots\}$				
Part of for loop	Code in C	Description	Structure of Java bytecode	
Initialization of	A = B	Value initializing	iconst, fconst, lconst, dconst, bi-	
an induction		an induction vari-	push, sipush, iload, fload, lload,	
variable		able	dload, getstatic, invokestatic, aload-	
			arraylength, aload-getfield, aload-	
			invokevirtual, aload-invokespecial,	
			aload-invokeinterface	
		store	istore, fstore, lstore, dstore	
Comparison part	A < C	Load an induction	Iload, fload, lload, dload	
		variable		
		Variable com-	iconst, fconst, lconst, dconst, bi-	
		pared with an	push, sipush, iload, fload, lload,	
		induction variable	dload, getstatic, invokestatic, aload-	
			arraylength, aload-getfield, aload-	
			invokevirtual, aload-invokespecial,	
			aload-invokeinterface	
		Conditional	If_icmpeq, if_icmpne, if_icmpge,	
			if_icmpgt, if_icmple, if_icmplt	
Body	$\{\cdots\}$	Body of for loop	Various instructions can occur	
Modification of an	A = A + D	Load an induction	iload, fload, lload, dload	
induction variable		variable		
		Modification	add or other various kinds of arith-	
			metic and logic operations.	
			istore, fstore, lstore, dstore	
		induction variable		
	A++	Autoincrement	iinc	

Table 1. for loop patterns which our compiler can recognize

of modification of an induction variable is that the iinc instruction should be followed by goto instruction for a backward jump.

backward jump: When the state machine is in state q9, it can be said that all needed information about for loop is found. Therefore, if the next instruction is goto that makes a backward jump to the start of a code section of a comparison part, the state machine goes into a final state. In order to justify a correct backward jump, it is checked whether a target address of the conditional jump is the next instruction of the backward jump.

Nested loops: In order to detect nested loops efficiently, we used a loop stack as a data structure. Whenever the state machine reaches to state q7 (after detecting an induction variable and a loop bound), all collected information is pushed into the loop stack and the state machine returns to state q0. In this way, our state machine can find all the front parts of loop patterns we define while scanning the whole code once. When reaching to the end of a whole code, the stack has the information about front parts of all the loops we found. The state machine pops from the stack one by one and returns to state q7. The pc also goes back

to the address of the conditional jump given by the popped information and the state machine starts checking whether those front parts of the loops have proper instructions for modification of an induction variables and backward jump.

4.3 Code Emission

Figure 4 (a) shows the Java source code and Java bytecode of an example of for loop patterns accepted by our state machine. And Figure 4 (b) shows our compiler's code generation without recognizing Java for loops.

(a) Java source code and bytecode		(b) Code generation using test and jump.	(c) for in C generated from for
for(11 = 8; 1 k+= 11;	l1 <i; l1++)<="" td=""><td>i0 = 8; Lvi10 = i0;</td><td>i0 = 8; Lvi10 = i0;</td></i;>	i0 = 8; Lvi10 = i0;	i0 = 8; Lvi10 = i0;
50.1: 1		_L56:	_L56:
52:bipush	8	i0 = Lvi10;	for(;
54:istore	10	i1 = Lvi3;	Lvi10 < Lvi3 ;
56:iload	10	if(i0 >= i1) goto _L75	•
58:iload_3		L62:	L62:
59:icmpge	75	i0 = Lvi4;	i0 = Lvi4;
62:iload	4	i1 = Lvi10;	i1 = Lvi10;
64:iload	10	i0 = i0 + i1;	•
66:iadd		Lvi4 = i0;	i0 = i0 + i1;
67:istore	4	•	Lvi4 = i0;
		Lvi10 += 1;	}
69:iinc		goto _L56;	
72:goto	56		

Fig. 4. An example of for loop

Our compiler is able to recognize a for loop in Java bytecodes and translates it into for in C. Figure 4 (c) presents the result obtained by applying this optimization technique to Java bytecode in Figure 4 (a).

For our experiment we implemented a simple case in Table 1. Especially, when a member field or a method of an object is used in code sections of initialization of an induction variable and a comparison part, translating for in Java into for in C is complicated due to an exception handling. Thus, we selected simple for loop patterns to be translated into for in C. Moreover, these patterns are the most frequently used patterns in codes. So, before generating for in C for all the for in Java, we checked the performance improvement of translating these patterns into for in C.

5 Performance Evaluation

5.1 Methodology

The performance of our Java-to-C compiler has been tested using Java Sci-Mark 2.0 benchmarks [14] on Zeon 2.0 GHz processor and 256 MB of memory

with Redhat Linux 8.0, and compared with gcj. The generated C codes were compiled by gcc 3.2 C compiler. The SciMark [14] is a composite Java benchmark measuring the performance of numerical codes occurring in scientific and engineering applications. The benchmark consists of the following five applications: Fast Fourier Transform, Gauss-Seidel relaxation, Sparse matrix-multiply, Monte Carlo integration, and dense LU factorization. Table 2 summarizes the applications.

Application	Description		
FFT	Fast Fourier Transform exercises complex arithmetic, shuffling, non-		
	constant memory references and trigonometric functions.		
SOR	Jacobi Successive Over-relaxation exercises typical accesses patterns in		
	finite difference applications, for example, solving Laplace's equation in		
	2D with Drichlet boundary conditions.		
Monte Carlo	Monte Carlo integration exercises random-number generators, synchro-		
	nized function calls, and function inlining.		
SparseMM	Sparse matrix multiply exercises indirection addressing and non-regular		
	memory references.		
LU	dense LU matrix factorization exercises linear algebra kernels (BLAS)		
	and dense matrix operations.		

Table 2. Java SciMark 2.0 Benchmark

5.2 Performance Comparison

The relative speedup to gcj is shown in Figure 5. Our Java-to-C compiler shows lower performance than gcj for all applications except SOR. Especially, in the case of Monte Carlo Integration, the overhead for synchronization in a singlethreaded application makes that our system has worse performance than gcj. In our compiler, even if an application is single-threaded, a monitor locking is enabled. Toba [8] can reduce the synchronization overhead, since the actual monitor locking is delayed until more than one thread are created. However the execution time is greatly improved by turning-off system-defined exception handling. The gci compiler optimizes an exception handling by using aggressive optimizations. Moreover, by translating for in Java into for in C, we can get faster execution time and see the possibility of further improvement in speedup. We could achieve 6% in SparseMM and 4% in Monte Carlo more speedup in loop code generation than test and jump. Such a little improvement is due to the limitation of loop patterns to be translated into for in C in our experiment. Our Java-to-C compiler generates for in C when modification of an induction variable is performed only by iinc statement. However for loop recognized by our compiler and translated into for in C is not the critical loop for performance in FFT, SOR and LU applications.

In the case of FFT, we could get much better performance when all major for loops are changed into for in C. For the experiment we found out major loops of each benchmark program, and then translated all the Java for loops

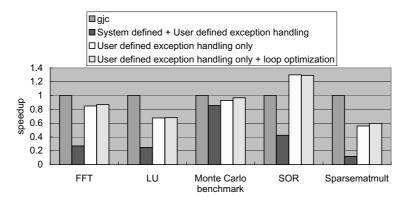


Fig. 5. Java SciMark 2.0 speedup

in for C by hand if our compiler did not translate them. The most critical method of FFT is transform_internal and it has four for loops. Our Java-to-C compiler translates one of them into for in C. When all the for loops are translated properly, the performance gets better up to 24% than test and jump code generation, and 5% faster than that gcj.

In the case of LU, the result is somewhat different. Although all the for loops in the most critical method of LU, factor were changed into for in C by hand, the performance was not improved. Therefore, we conclude that for loops in factor of LU are not an important factor in performance and loop optimization provided by gcc compiler is not very helpful in improving the performance. In the case of SOR, we found out that all the loops in the most critical method, execute, are translated in for in C already by our compiler. Therefore, we can conclude that in SOR, for loops are not the determining factor in overall performance.

6 Conclusion

In this paper, we presented the structure of the Java-to-C compiler to preserve the Java semantics, and discuss code generation and optimization issues. we presented the implementation of translating for in Java into for in C in order to make performance better. Because of limitation of loop patterns that can be translated for in C, we could not get as high performance as expected. However, by changing the rest for loops into for in C, we can discover the possibility of further improvement.

The generated C codes include many pointer and complex expressions, which prevent AOT compilers from applying advanced compiler optimization techniques like constant propagation, sub-expression elimination, inlining, and so on. In ongoing research, we have developed an IR framework between bytecode and C codes for helping AOT compilers generate better quality of codes and also have made our compiler generate for in C for more broad range of for loop patterns in Java source code.

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