

Advanced Compiler Construction Theory And Practice



Introduction to loop dependence
and Optimizations

A little about myself

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Research Interests

- Compiler construction, software productivity
- program analysis and optimization for high-performance computing
- Parallel programming

Overall goal: develop tools to improve the productivity and efficiency of programming

- Optimizing compilers for high performance computing
- Programmable optimization and tuning of Scientific codes

General Information

□ Reference books

- Optimizing Compilers for Modern Architectures: A Dependence-based Approach,
 - Randy Allen and Ken Kennedy, Morgan-Kaufman Publishers Inc.
- Book Chapter: Optimizing And Tuning Scientific Codes
 - Qing Yi. SCALABLE COMPUTING AND COMMUNICATIONS:THEORY AND PRACTICE.
 - <http://www.cs.uccs.edu/~qyi/papers/BookChapter11.pdf>
- Structured Parallel Programming Patterns for Efficient Computation
 - Michael McCool, James Reinders, and Arch Robison. Morgan Kaufmann. 2012.
 - http://parallelbook.com/sites/parallelbook.com/files/SC13_20131117_Intel_McCool_Robison_Reinders_Hebenstreit.pdf

□ Course materials

- <http://www.cs.uccs.edu/~qyi/classes/Dragonstar>

□ The POET project web site:

- <http://www.cs.uccs.edu/~qyi/poet/index.php>

High Performance Computing

- Applications must efficiently manage architectural components
 - Parallel processing
 - Multiple execution units --- pipelined
 - Vector operations, multi-core
 - Multi-tasking, multi-threading (SIMD), and message-passing
 - Data management
 - Registers
 - Cache hierarchy
 - Shared and distributed memory
 - Combinations of the above

- What are the compilation challenges?

Optimizing For High Performance

- Goal: eliminating inefficiencies in programs
- Eliminating redundancy: if an operation has already been evaluated, don't do it again
 - Especially if the operation is inside loops or part of a recursive evaluation
 - All optimizing compilers apply redundancy elimination,
 - e.g., loop invariant code motion, value numbering, global RE
- Resource management: reorder operations and/or data to better map to the targeting machine
 - Reorder computation(operations)
 - parallelization, vectorization, pipelining, VLIW, memory reuse
 - Instruction scheduling and loop transformations
 - Re-organization of data
 - Register allocation, regrouping of arrays and data structures

Optimizing For Modern Architectures

- Key: reorder operations to better manage resources
 - Parallelization and vectorization
 - memory hierarchy management
 - Instruction and task/threads scheduling
 - Interprocedural (whole-program) optimizations
- Most compilers focus on optimizing loops, why?
 - This is where the application spends most of its computing time
 - What about recursive function/procedural calls?
 - Extremely important, but often left unoptimized...

Compiler Technologies

- Source-level optimizations
 - Most architectural issues can be dealt with by restructuring the program source
 - Vectorization, parallelization, data locality enhancement
 - Challenges:
 - Determining when optimizations are legal
 - Selecting optimizations based on profitability
- Assembly level optimizations
 - Some issues must be dealt with at a lower level
 - Prefetch insertion
 - Instruction scheduling
- All require some understanding of the ways that instructions and statements depend on one another (share data)

Syllabus

- Dependence Theory and Practice
 - Automatic detection of parallelism.
 - Types of dependences; Testing for dependence;
- Memory Hierarchy Management
 - Locality enhancement; data layout management;
 - Loop interchange, blocking, unroll&jam, unrolling,...
- Loop Parallelization
 - More loop optimizations: OMP parallelization, skewing, fusion, ...
 - Private and reduction variables
- Pattern-driven optimization
 - Structured Parallelization Patterns
 - Pattern-driven composition of optimizations
- Programmable optimization and tuning
 - Using POET to write your own optimizations

Dependence-based Optimization

□ **Bernstein's Conditions**

- **it is safe to run two tasks R1 and R2 in parallel if none of the following holds:**
 - **R1 writes into a memory location that R2 reads**
 - **R2 writes into a memory location that R1 reads**
 - **Both R1 and R2 write to the same memory location**

- There is a dependence between two statements if
 - They might access the same location,
 - There is a path from one to the other, and
 - One of the accesses is a write
- Dependence can be used for
 - Automatic parallelization
 - Memory hierarchy management (registers and caches)
 - Scheduling of instructions and tasks

Dependence - Static Program Analysis

- Program analysis --- support software development and maintenance
 - Compilation --- identify errors without running program
 - Smart development environment (check simple errors as you type)
 - Optimization --- cannot not change program meaning
 - Improve performance, efficiency of resource utilization
 - Code revision/re-factoring ==> reusability, maintainability,
 - Program correctness --- Is the program safe/correct?
 - Program verification -- Is the implementation safe, secure?
 - Program integration --- are there any communication errors?
- In contrast, if the program needs to be run to figure out information, it is called dynamic program analysis.

Data Dependences

- **There is a data dependence from statement S_1 to S_2 if**
 1. **Both statements access the same memory location,**
 2. **At least one of them stores onto it, and**
 3. **There is a feasible run-time execution path from S_1 to S_2**
- Classification of data dependence
 - True dependences (Read After Write hazard)
 S_2 depends on S_1 is denoted by $S_1 \delta S_2$
 - Anti dependence (Write After Read hazard)
 S_2 depends on S_1 is denoted by $S_1 \delta^{-1} S_2$
 - Output dependence (Write After Write hazard)
 S_2 depends on S_1 is denoted by $S_1 \delta^0 S_2$
- Simple example of data dependence:
 S_1 PI = 3.14
 S_2 R = 5.0
 S_3 AREA = PI * R ** 2

Transformations

- A reordering Transformation
 - Changes the execution order of the code, without adding or deleting any operations.
- Properties of Reordering Transformations
 - It does not eliminate dependences, but can change the ordering (relative source and sink) of a dependence
 - If a dependence is reverted by a reordering transformation, it may lead to incorrect behavior
- A reordering transformation is safe if it preserves the relative direction (i.e., the source and sink) of each dependence.

Dependence in Loops

```
DO I = 1, N
S1  A(I+1) = A(I) + B(I)
ENDDO
```

```
DO I = 1, N
S1  A(I+2) = A(I) + B(I)
ENDDO
```

- In both cases, statement S1 depends on itself
 - However, there is a significant difference
- We need to distinguish different iterations of loops
 - The iteration number of a loop is equal to the value of the loop index (loop induction variable)
 - Example:

```
DO I = 0, 10, 2
S1  <some statement>
ENDDO
```
- What about nested loops?
 - Need to consider the nesting level of a loop

Iteration Vectors

- Given a nest of n loops, iteration vector i is
 - A vector of integers $\{i_1, i_2, \dots, i_n\}$
where $i_k, 1 \leq k \leq n$ represents the iteration number for the loop at nesting level k

- Example:

```
DO I = 1, 2
  DO J = 1, 2
    S1      <some statement>
  ENDDO
ENDDO
```

- The iteration vector $(2, 1)$ denotes the instance of S_1 executed during the 2nd iteration of the I loop and the 1st iteration of the J loop

Loop Iteration Space

- For each loop nest, its iteration space is
 - The set of all possible iteration vectors for a statement
 - Example:

```
DO I = 1, 2  
  DO J = 1, 2  
    S1    <some statement>  
  ENDDO  
ENDDO
```

The iteration space for S1 is $\{ (1,1), (1,2), (2,1), (2,2) \}$

Ordering of Iterations

- Within a single loop iteration space,
 - We can impose a lexicographic ordering among its iteration Vectors
- Iterations i precedes j , denoted $i < j$, iff
 - Iteration i is evaluated before j
 - That is, for some nesting level k
 1. $i[i:k-1] < j[1:k-1]$, or
 2. $i[1:k-1] = j[1:k-1]$ and $i_n < j_n$
 - Example: $(1,1) < (1,2) < (2,1) < (2,2)$

Loop Dependence

There exists a dependence from statement S1 to S2 in a common nest of loops if and only if

- there exist two iteration vectors i and j for the nest, such that
 - (1) $i < j$ or $i = j$ and there is a path from S1 to S2 in the body of the loop,
 - (2) statement S1 accesses memory location M on iteration i and statement S2 accesses location M on iteration j , and
 - (3) one of these accesses is a write.

Distance and Direction Vectors

- Consider a dependence in a loop nest of n loops
 - Statement $S1$ on iteration i is the source of dependence
 - Statement $S2$ on iteration j is the sink of dependence
- The distance vector is a vector of length n $d(i,j)$ such that:
 $d(i,j)_k = j_k - I_k$
- The direction Vector is a vector of length n $D(i,j)$ such that
(Definition 2.10 in the book)

$$D(i,j)_k = \begin{cases} "<" & \text{if } d(i,j)_k > 0 \\ "=" & \text{if } d(i,j)_k = 0 \\ ">" & \text{if } d(i,j)_k < 0 \end{cases}$$

- What is the dependence distance/direction vector?

```
DO I = 1, N
  DO J = 1, M
    DO K = 1, L
      S1    A(I+1, J, K-1) = A(I, J, K) + 10
```

Loop-carried and Loop-independent Dependences

- If in a loop statement S2 on iteration j depends on S1 on iteration i , the dependence is
 - **Loop-carried** if either of the following conditions is satisfied
 - S1 and S2 execute on different iterations i.e., $i \neq j$
 - $d(i,j) > \mathbf{0}$ i.e. $D(i,j)$ contains a "<" as leftmost non "=" component
 - **Loop-independent** if either of the conditions is satisfied
 - S1 and S2 execute on the same iteration i.e., $i=j$
 - $d(i,j) = \mathbf{0}$, i.e. $D(i,j)$ contains only "=" component
 - NOTE: there must be a path from S1 to S2 in the same iteration

- Example:

```
DO I = 1, N
S1      A(I+1) = F(I) + A(I)
S2      F(I) = A(I+1)
ENDDO
```

Level of loop dependence

- The level of a loop-carried dependence is the index of the leftmost non-“=” of $D(i,j)$
 - A level- k dependence from S_1 to S_2 is denoted $S_1 \delta_k S_2$
 - A loop independent dependence from S_1 to S_2 is denoted $S_1 \delta_\infty S_2$
- Example:

```
DO I = 1, 10
  DO J = 1, 10
    DO K = 1, 10
      S1      A(I, J, K+1) = A(I, J, K)
      S2      F(I, J, K) = A(I, J, K+1)
    ENDDO
  ENDDO
ENDDO
```

Loop Optimizations

- A loop optimization may
 - Change the order of iterating the iteration space of each statement
 - Without altering the direction of any original dependence
- Example loop transformations
 - Change the nesting order of loops
 - Fuse multiple loops into one or split one into multiple
 - Change the enumeration order of each loop
 - And more ...
- Any loop reordering transformation that
 - (1) does not alter the relative nesting order of loops and
 - (2) preserves the iteration order of the level-k looppreserves all level-k dependences.

Dependence Testing

```
DO i1 = L1, U1, S1
  DO i2 = L2, U2, S2
    ...
    DO in = Ln, Un, Sn
      S1   A(f1(i1,...,in),...,fm(i1,...,in)) = ...
      S2   ... = A(g1(i1,...,in),...,gm(i1,...,in))
    ENDDO
  ENDDO
ENDDO
```

- A dependence exists from S1 to S2 iff there exist iteration vectors $x=(x_1,x_2,\dots,x_n)$ and $y=(y_1,y_2,\dots,y_n)$ such that
 - (1) x is lexicographically less than or equal to y ;
 - (2) the system of **diophantine equations** has an integer solution:
$$f_i(x) = g_i(y) \text{ for all } 1 \leq i \leq m$$
i.e. $f_i(x_1,\dots,x_n)-g_i(y_1,\dots,y_n)=0$ for all $1 \leq i \leq m$

Example

```
DO I = 1, 10
  DO J = 1, 10
    DO K = 1, 10
      S1      A(I, J, K+1) = A(I, J, K)
      S2      F(I, J, K) = A(I, J, K+1)
    ENDDO
  ENDDO
ENDDO
```

- To determine the dependence between $A(I, J, K+1)$ at iteration vector (I_1, J_1, K_1) and $A(I, J, K)$ at iteration vector (I_2, J_2, K_2) , solve the system of equations
 - $I_1 = I_2; J_1 = J_2; K_1 + 1 = K_2; 1 \leq I_1, I_2, J_1, J_2, K_1, K_2 \leq 10$
 - Distance vector is $(I_2 - I_1, J_2 - J_1, K_2 - K_1) = (0, 0, 1)$
 - Direction vector is $(=, =, <)$
 - The dependence is from $A(I, J, K+1)$ to $A(I, J, K)$ and is a true dependence

The Delta Notation

- Goal: compute iteration distance between the source and sink of a dependence

```
DO I = 1, N
  A(I + 1) = A(I) + B
ENDDO
```

- Iteration at source/sink denoted by: I_0 and $I_0 + \Delta I$
- Forming an equality gets us: $I_0 + 1 = I_0 + \Delta I$
- Solving this gives us: $\Delta I = 1$
- If a loop index does not appear, its distance is *

- * means the union of all three directions $<, >, =$

```
DO I = 1, 100
  DO J = 1, 100
    A(I+1) = A(I) + B(J)
```

- The direction vector for the dependence is $(<, *)$

Complexity of Testing

- Find integer solutions to a system of Diophantine Equations is NP-Complete
 - Most methods consider only linear subscript expressions
- Conservative Testing
 - Try to prove absence of solutions for the dependence equations
 - Conservative, but never incorrect
- Categorizing subscript testing equations
 - ZIV if it contains no loop index variable
 - SIV if it contains only one loop index variable
 - MIV if it contains more than one loop index variables

$$A(5, I+1, j) = A(1, I, k) + C$$

5=1 is ZIV; I1+1=I2 is SIV; J1=K2 is MIV

Summary

- Introducing data dependence
 - What is the meaning of S2 depends on S1?
 - What is the meaning of $S_1 \delta S_2$, $S_1 \delta^{-1} S_2$, $S_1 \delta^0 S_2$?
 - What is the safety constraint of reordering transformations?
- Loop dependence
 - What is the meaning of iteration vector (3,5,7)?
 - What is the iteration space of a loop nest?
 - What is the meaning of iteration vector $I < J$?
 - What is the distance/direction vector of a loop dependence?
 - What is the relation between dependence distance and direction?
 - What is the safety constraint of loop reordering transformations?
- Level of loop dependence and transformations
 - What is the meaning of loop carried/independent dependences?
 - What is the level of a loop dependence or loop transformation?
 - What is the safety constraint of loop parallelization?
- Dependence testing theory