

Autotuning

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What is Autotuning?

- Searching for the "best" code parameters, code transformations, system configuration settings, etc.
- Search can be
 - Quasi-intelligent: genetic algorithms, hill-climbing
 - Random (often quite good!)



Parameters to tune in all of these

application

compiler

runtime system

operating system

virtualization

hardware

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Traditional Compilers

- "One size fits all" approach
- Tuned for average performance
- Aggressive opts often turned off
- Target hard to model analytically

application
compiler
runtime system
operating system
virtualization
hardware



Solution : Random Search!

- Identify large set of optimizations to search over
 - Some optimizations require parameter values, search over those values also!
 - Out-performs state-of-the-art compiler





Optimization Sequence Representation

Use random number generator to construct sequence

Example:

-LNO:interchange=1:prefetch=2:blocking_size=32:fusion=1:... Generate each of these parameter values using a random number generator Note: need to define a range of interesting values a-priori

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Case Study: Random vs State-of-the-Art

- PathScale compiler
 - Compare to highest optimization level
 - 121 compiler flags
- AMD Athlon processor
 - *Real* machine; Not simulation
- 57 benchmarks
 - SPEC (INT 95, INT/FP 2000), MiBench, Polyhedral



Evaluated Search Strategies

- RAND
 - Randomly select 500 optimization sequences

Combined Elimination [CGO 2006]

- Pure search technique
 - Evaluate optimizations one at a time
 - Eliminate negative optimizations in one go
- Out-performed other pure search techniques
- PC Model [CGO 2007]
 - Machine learning model using performance counters
 - Mapping of performance counters to good optimizations

Performance vs Evaluations



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Some recommendations

- Use small input sizes that are representative
 - Be careful as tuning on small inputs may not give you the best performance on regular (larger) inputs
- Reduce application to most important kernel(s) and tune those
- If kernels can be mapped to highly-tuned library implementations, use those!



Some recommendations

- No optimization search will help a bad algorithm!
 - Chose the correct algorithm first!

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Using quasi-intelligent search

- Can easily setup a genetic algorithm or hill-climbing to perform search over optimization space
- We can help you set this up.
- Random often does as well!



Performance counters

- Can be used to narrow search to particular set of optimizations
- Lots of cache misses may require loop restructuring, e.g., blocking
- Lots of resource stalls may require instruction scheduling



Most Informative Perf Counters

- 1. L1 Cache Accesses
- 2. L1 Dcache Hits
- **3. TLB Data Misses**
- **4. Branch Instructions**
- **5. Resource Stalls**
- 6. Total Cycles
- 7. L2 Icache Hits
- 8. Vector Instructions

- 9. L2 Dcache Hits
- **10. L2 Cache Accesses**
- 11. L1 Dcache Accesses
- **12. Hardware Interrupts**
- 13. L2 Cache Hits
- 14. L1 Cache Hits
- **15. Branch Misses**



Dependence Analysis and Loop Transformations

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Lecture Overview

- Very Brief Introduction to Dependences
- Loop Transformations



The Big Picture

What are our goals?

Simple Goal: Make execution time as small as possible

Which leads to:

- Achieve execution of many (all, in the best case) instructions in parallel
- Find <u>independent</u> instructions



Dependences

- We will concentrate on data dependences
- Simple example of data dependence:

$$S_1 PI = 3.14$$

$$S_2 R = 5.0$$

- S_3 AREA = PI * R ** 2
- Statement S₃ cannot be moved before either
 S₁ or S₂ without compromising correct results



Dependences

- Formally:
 - There is a data dependence from statement S_1 to statement S_2 (S_2 depends on S_1) if:
 - 1. Both statements access the same memory location and at least one of them stores onto it, and
 - 2. There is a feasible run-time execution path from $\ensuremath{S_1}$ to $\ensuremath{S_2}$



Load Store Classification

- Quick review of dependences classified in terms of load-store order:
 - 1. True dependence (RAW hazard)
 - 2. Antidependence (WAR hazard)
 - 3. Output dependence (WAW hazard)



Dependence in Loops

• Let us look at two different loops:

DO I = 1, N $S_1 \quad A(I+1) = A(I) + B(I)$ ENDDO DO I = 1, N $S_1 \quad A(I+2) = A(I)+B(I)$ ENDDO

• In both cases, statement S₁ depends on itself



Transformations

- We call a transformation safe if the transformed program has the same "meaning" as the original program
- But, what is the "meaning" of a program?
- For our purposes:
- Two computations are equivalent if, on the same inputs:
 - They produce the same outputs in the same order



Reordering Transformations

 Is any program transformation that changes the order of execution of the code, without adding or deleting any executions of any statements



Properties of Reordering Transformations

- A reordering transformation does not eliminate dependences
- However, it can change the ordering of the dependence which will lead to incorrect behavior
- A reordering transformation preserves a dependence if it preserves the relative execution order of the source and sink of that dependence.



Loop Transformations

- Compilers have always focused on loops
 - Higher execution counts
 - Repeated, related operations
- Much of real work takes place in loops



Several effects to attack

- Overhead
 - Decrease control-structure cost per iteration
- Locality
 - Spatial locality \Rightarrow use of co-resident data
 - Temporal locality \Rightarrow reuse of same data
- Parallelism
 - Execute independent iterations of loop in parallel



Eliminating Overhead

Loop unrolling (the oldest trick in the book)To reduce overhead, replicate the loop body

- do i = 1 to 100 by 1 a(i) = a(i) + b(i)end (unroll by 4)
- do i = 1 to 100 by 4 a(i) = a(i) + b(i) a(i+1) = a(i+1) + b(i+1) a(i+2) = a(i+2) + b(i+2) a(i+3) = a(i+3) + b(i+3)end

Sources of Improvement

- Less overhead per useful operation
- Longer basic blocks for local optimization



Eliminating Overhead

 Loop unrolling with unknown bounds
 Generate guard loops i=1 do while (i+3 < n)

do i = 1 to n by 1 a(i) = a(i) + b(i) end *becomes* (unroll by 4)

a(i) = a(i) + b(i) a(i+1) = a(i+1) + b(i+1) a(i+2) = a(i+2) + b(i+2) a(i+3) = a(i+3) + b(i+3) i = i + 4end



Eliminating Overhead

One other use for loop unrollingEliminate copies at the end of a loop

$$\begin{array}{ll} t_1 = b(0) \\ do \ i = 1 \ to \ 100 \\ a(i) = a(i) + t_1 + t_2 \\ t_1 = t_2 \\ end \end{array} \begin{array}{ll} t_1 = b(0) \\ becomes \\ (unroll + rename) \\ t_1 = b(i) \\ a(i) = a(i) + t_1 + t_2 \\ t_1 = b(i) \\ a(i) = a(i) + t_1 + t_2 \\ t_1 = b(i+1) \\ a(i+1) = a(i+1) + t_2 + t_4 \\ end \end{array}$$



Loop Unswitching

- Hoist invariant control-flow out of loop nest
- Replicate the loop & specialize it
- No tests, branches in loop body
- Longer segments of straight-line code



Loop Unswitching

loon		If test then
	Іоор	
statements		statements
if test then	<i>becomes</i> (unswitch)	then part
then part		more statements
else		endloop
else part		else
endif		Іоор
more statements		statements
endloop		else part
		more statements
		endloop
		endif *

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Loop Unswitching

do i = 1 to 100 a(i) = a(i) + b(i) if (expression) then d(i) = 0 end

becomes

```
(unswitch)
```

```
if (expression) then
do i = 1 to 100
a(i) = a(i) + b(i)
d(i) = 0
end
else
do i = 1 to 100
a(i) = a(i) + b(i)
end
```



Loop Fusion

- Two loops over same iteration space \Rightarrow one loop
- Safe if does not change the values used or defined by any statement in either loop (i.e., does not violate deps)

$$\begin{array}{lll} \mbox{do i = 1$ to n} \\ \mbox{c(i) = a(i) + b(i)$} \\ \mbox{end} \\ \mbox{do j = 1$ to n} \\ \mbox{d(j) = a(j) * e(j)$} \\ \mbox{end} \end{array} \qquad \begin{array}{lll} \mbox{do i = 1$ to n} \\ \mbox{c(i) = a(i) + b(i)$} \\ \mbox{d(i) = a(i) * e(i)$} \\ \mbox{end} \\ \mbox{end} \end{array}$$

For big arrays, a(i) may not be in the cache

a(i) will be found in the cache



Loop Fusion Advantages

- Enhance temporal locality
- Reduce control overhead
- Longer blocks for local optimization & scheduling
- Can convert inter-loop reuse to intra-loop reuse



Loop Fusion of Parallel Loops

- Parallel loop fusion legal if dependences loop independent
 - Source and target of flow dependence map to same loop iteration



- Single loop with independent statements \Rightarrow multiple loops
- Starts by constructing statement level dependence graph
- Safe to perform distribution if:
 - No cycles in the dependence graph
 - Statements forming cycle in dependence graph put in same loop







- (1) for I = 1 to N do
- (2) A[I] = A[i] + B[i-1]
- (3) B[I] = C[I-1]*X+C
- (4) C[I] = 1/B[I]
- (5) D[I] = sqrt(C[I])
- (6) endfor

Has the following dependence graph





- (1) for I = 1 to N do
- (2) A[I] = A[i] + B[i-1]
- (3) B[I] = C[I-1]*X+C
- (4) C[I] = 1/B[I]
- (5) D[I] = sqrt(C[I])
- (6) endfor

becomes

(fission)

(1) for I = 1 to N do (2) A[I] = A[I] + B[I-1](3) endfor (4) for (5) B[I] = C[I-1]*X+C(6) C[I] = 1/B[I](7)endfor (8)for (9) D[I] = sqrt(C[I])(10)endfor



Loop Fission Advantages

- Enables other transformations
 - E.g., Vectorization
- Resulting loops have smaller cache footprints
 - More reuse hits in the cache



Loop Interchange

- do i = 1 to 50 do j = 1 to 100 a(i,j) = b(i,j) * c(i,j) (interchange) end end do j = 1 to 100 do i = 1 to 50 a(i,j) = b(i,j) * c(i,j)end end end end end
 - Swap inner & outer loops to rearrange iteration space
 Effect
 - Improves reuse by using more elements per cache line
 - Goal is to get as much reuse into inner loop as possible



Loop Interchange Effect

- If one loop carries all dependence relations
 - Swap to outermost loop and all inner loops executed in parallel
- If outer loops iterates many times and inner only a few
 - Swap outer and inner loops to reduce startup overhead
- Improves reuse by using more elements per cache line
- Goal is to get as much reuse into inner loop as possible



Reordering Loops for Locality

In <u>row-major</u> order, the opposite loop ordering causes the same effects

In Fortran's column-major order, a(4,4) would lay out as



After interchange, direction of Iteration is changed



cache line

As little as 1 used element per line

Runs down cache line



Loop permutation

- Interchange is degenerate case
 - Two perfectly nested loops
- More general problem is called permutation

Safety

- Permutation is safe <u>iff</u> no data dependences are reversed
 - The flow of data from definitions to uses is preserved



Loop Permutation Effects

- Change order of access & order of computation
- Move accesses closer in time ⇒ increase temporal locality
- Move computations farther apart ⇒ cover pipeline latencies



Strip Mining

Splits a loop into two loops

```
do j = 1 to 100
do i = 1 to 50
a(i,j) = b(i,j) *
endend
do j = 1 to 100
do ii = 1 to 50 by 8
do i = ii to min(ii+7,50)
a(i,j) = b(i,j) * c(i,j)
end
end
end
```

Note: This is always safe, but used by itself not profitable!

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Strip Mining Effects

- May slow down the code (extra loop)
- Enables vectorization



```
do t = 1,T
do i = 1,n
do j = 1,n
... a(i,j) ...
end do
end do
end do
```

Want to exploit temporal locality in loop nest.





B: Block Size



```
do ic = 1, n, B

do jc = 1, n, B

do t = 1,T

do i = ic, min(n,ic+B-1), 1

do j = jc, min(n, jc+B-1), 1

... a(i,j) ...

end do

end do

end do

end do

end do

end do

end do
```



B: Block Size



```
do ic = 1, n, B

do jc = 1, n, B

do t = 1,T

do i = ic, min(n,ic+B-1), 1

do j = jc, min(n, jc+B-1), 1

... a(i,j) ...

end do

end do
```

B: Block Size



```
do ic = 1, n, B

do jc = 1, n, B

do t = 1,T

do i = ic, min(n,ic+B-1), 1

do j = jc, min(n, jc+B-1), 1

... a(i,j) ...

end do

end do

end do

end do

end do

end do
```



B: Block Size When is this legal?

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Loop Tiling Effects

- Reduces volume of data between reuses
 - Works on one "tile" at a time (tile size is B by B)
- Choice of tile size is crucial



Scalar Replacement

- Allocators never keep c(i) in a register
- We can trick the allocator by rewriting the references

The plan

- Locate patterns of consistent reuse
- Make loads and stores use temporary scalar variable
- Replace references with temporary's name



Scalar Replacement



Almost any register allocator can get t into a register



Scalar Replacement Effects

- Decreases number of loads and stores
- Keeps reused values in names that can be allocated to registers
- In essence, this exposes the reuse of a(i) to subsequent passes