

Performance Improvement Revisited

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Goals of this Lecture



- Help you learn how to:
 - Improve program performance by exploiting knowledge of underlying system
 - · Compiler capabilities
 - · Hardware architecture
 - Program execution
- And thereby:
 - · Help you to write efficient programs
 - · Review material from the second half of the course

Improving Program Performance



- Most programs are already "fast enough"
 - · No need to optimize performance at all
 - Save your time, and keep the program simple/readable
- Most parts of a program are already "fast enough"
 - Usually only a small part makes the program run slowly
 - · Optimize only this portion of the program, as needed
- · Steps to improve execution (time) efficiency
 - Do timing studies (e.g., gprof)
 - · Identify hot spots
 - · Optimize that part of the program
 - · Repeat as needed

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Ways to Optimize Performance



- Better data structures and algorithms
- Improves the "asymptotic complexity"
 - Better scaling of computation/storage as input grows
 - E.g., going from $O(n^2)$ sorting algorithm to $O(n \log n)$
 - Clearly important if large inputs are expected
 - Requires understanding data structures and algorithms
- Better source code the compiler can optimize
 - Improves the "constant factors"
 - Faster computation during each iteration of a loop
 - E.g., going from 1000n to 10n running time
 - Clearly important if a portion of code is running slowly
 - · Requires understanding hardware, compiler, execution

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Helping the Compiler Do Its Job

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Optimizing Compilers



- Provide efficient mapping of program to machine
 - Register allocation
 - · Code selection and ordering
 - · Eliminating minor inefficiencies
- Don't (usually) improve asymptotic efficiency
 - Up to the programmer to select best overall algorithm
- Have difficulty overcoming "optimization blockers"
 - Potential function side-effects
 - · Potential memory aliasing

Limitations of Optimizing Compilers 7



- Fundamental constraint
 - · Compiler must not change program behavior
 - · Ever, even under rare pathological inputs
- Behavior that may be obvious to the programmer can be obfuscated by languages and coding styles
 - Data ranges more limited than variable types suggest
 - Array elements remain unchanged by function calls
- Most analysis is performed only within functions
 - Whole-program analysis is too expensive in most cases
- Most analysis is based only on static information
 - Compiler has difficulty anticipating run-time inputs

Avoiding Repeated Computation



- A good compiler recognizes simple optimizations
 - Avoiding redundant computations in simple loops
 - Still, programmer may still want to make it explicit
- Example
 - Repetition of computation: n * i

```
for (i = 0; i < n; i++)
 for (j = 0; j < n; j++)
    a[n*i + j] = b[j];
```

```
for (i = 0; i < n; i++) {
 int ni = n * i;
 for (j = 0; j < n; j++)
   a[ni + j] = b[j];
```

Worrying About Side Effects



Is this transformation okay?

```
int func1(int x) {
   return f(x) + f(x) + f(x) + f(x);
}
```

· Not necessarily, if

```
int counter = 0;
int f(int x) {
  return counter++;
}
```

int func1(int x) {
 return 4 * f(x);
}

And this function may be defined in another file known only at link time!

- Compiler cannot always avoid repeated computation
 - · May not know if the code has a "side effect"
 - · ... that makes the transformation change the code's behavior

Another Example on Side Effects



• Is this optimization okay for compiler to do?

```
for (i = 0; i < strlen(s); i++) {
    /* Do something with s[i] */
}</pre>
```

```
length = strlen(s);
for (i = 0; i < length; i++) {
    /* Do something with s[i] */
}</pre>
```

- Short answer: it depends
 - Compiler often cannot tell
 - Most compilers do not try to identify side effects
- Programmer knows best
 - · And can decide whether the optimization is safe

Memory Aliasing



• Is this optimization okay?

```
void twiddle(int *xp, int *yp) {
    *xp += *yp;
    *xp += *yp;
}
```

```
void twiddle(int *xp, int *yp) {
   *xp += 2 * *yp;
}
```

- Not necessarily, what if xp and yp are equal?
 - First version: result is 4 times *xp
 - Second version: result is 3 times *xp

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Memory Aliasing



- Memory aliasing
 - · Single data location accessed through multiple names
 - E.g., two pointers that point to the same memory location
- Modifying the data using one name
 - Implicitly modifies the values seen through other names

xp, yp →

- · Blocks optimization by the compiler
 - The compiler cannot tell when aliasing may occur
 - ... and so must forgo optimizing the code
- Programmer often does know
 - And can optimize the code accordingly

Another Aliasing Example



Is this optimization okay?

```
int *x, *y;
...
*x = 5;
*y = 10;
printf("x=%d\n", *x);

printf("x=5\n");
```

- Not necessarily
 - If y and x point to the same location in memory...
 - ... the correct output is "x = 10\n"

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Summary: Helping the Compiler



- Compiler can perform many optimizations
 - · Register allocation
 - · Code selection and ordering
 - · Eliminating minor inefficiencies
- But often the compiler needs your help
 - Knowing if code is free of side effects
 - Knowing if memory aliasing will not happen
- Modifying the code can lead to better performance
 - Profile the code to identify the "hot spots"
 - Look at the assembly language the compiler produces
 - · Rewrite the code to get the compiler to do the right thing



Exploiting the Hardware

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Underlying Hardware



- Implements a collection of instructions
- Instruction set varies from one architecture to another
 - · Some instructions may be faster than others
- Registers and caches are faster than main memory
 - · Number of registers and sizes of caches vary
 - · Exploiting both spatial and temporal locality
- Exploits opportunities for parallelism
 - · Pipelining: decoding one instruction while running another
 - · Benefits from code that runs in a sequence
 - Superscalar: perform multiple operations per clock cycle
 - · Benefits from operations that can run independently
 - Speculative execution: performing instructions before knowing they will be reached (e.g., without knowing outcome of a branch)

Addition Faster Than Multiplication



- Adding instead of multiplying
 - · Addition is faster than multiplication
- Recognize sequences in products
 - Replace multiplication with repeated addition

```
for (i = 0; i < n; i++) {
 int ni = n * i;
 for (j = 0; j < n; j++)
   a[ni + j] = b[j];
```

```
int ni = 0;
for (i = 0; i < n; i++) {
 for (j = 0; j < n; j++)
   a[ni + j] = b[j];
 ni += n;
```

Bit Operations Faster Than Arithmetic



- Use shifts to multiply/divide by powers of 2
 - "x >> 3" is faster than "x/8"
 - "x << 3" is faster than "x * 8"

53 | 0 | 0 | 1 | 1 | 0 | 1 | 0 | 1

53<<2 1 1 0 1 0 0 0 0

- Bit masking is faster than mod operation
 - "x & 15" is faster than "x % 16" 53 0 0 1 1 0

& 15 | 0 | 0 | 0 | 0

5 | 0 | 0 | 0 | 0 | 0 | 1 | 0 | 1

Caching: Matrix Multiplication



- Caches
 - Slower than registers, but faster than main memory
 - · Both instruction caches and data caches
- Locality
 - Temporal locality: recently-referenced items are likely to be referenced in near future
 - Spatial locality: Items with nearby addresses tend to be referenced close together in time
- Matrix multiplication
 - · Multiply n-by-n matrices A and B, and store in matrix C
 - Performance heavily depends on effective use of caches,

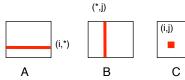
Matrix Multiply: Cache Effects



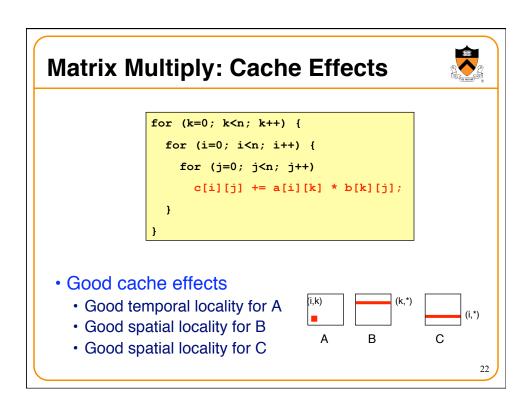
```
for (i=0; i<n; i++) {
  for (j=0; j<n; j++) {
    for (k=0; k<n; k++)
      c[i][j] += a[i][k] * b[k][j];
  }
}</pre>
```

- Reasonable cache effects
 - Good spatial locality for A
 - · Poor spatial locality for B

Good temporal locality for C



Matrix Multiply: Cache Effects for (j=0; j<n; j++) { for (k=0; k<n; k++) { for (i=0; i<n; i++) c[i][j] += a[i][k] * b[k][j]; } } • Rather poor cache effects • Bad spatial locality for A • Good temporal locality for B • Bad spatial locality for C



Parallelism: Loop Unrolling



· What limits the performance?

```
for (i = 0; i < length; i++)
sum += data[i];</pre>
```

- Limited apparent parallelism
 - One main operation per iteration (plus book-keeping)
 - Not enough work to keep multiple functional units busy
 - Disruption of instruction pipeline from frequent branches
- Solution: unroll the loop
 - Perform multiple operations on each iteration

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Parallelism: After Loop Unrolling



Original code

```
for (i = 0; i < length; i++)
sum += data[i];</pre>
```

After loop unrolling (by three)

```
/* Combine three elements at a time */
limit = length - 2;
for (i = 0; i < limit; i+=3)
   sum += data[i] + data[i+1] + data[i+2];

/* Finish any remaining elements */
for (; i < length; i++)
   sum += data[i];</pre>
```



Program Execution

Writing Your Own Malloc and Free



- Dynamic memory management
 - malloc() to allocate blocks of memory
 - free () to free blocks of memory
- Existing malloc() and free() implementations
 - Designed to handle a wide range of reguest sizes
 - · Good most of the time, but rarely the best for all workloads
- Designing your own dynamic memory management
 - Forego using traditional malloc() and free(), and write your own
 - · E.g., if you know all blocks will be the same size
 - · E.g., if you know blocks will usually be freed in the order allocated
 - E.g., <insert your known special property here>

Conclusion



- Work smarter, not harder
 - No need to optimize a program that is "fast enough"
 - Optimize only when, and where, necessary
- Speeding up a program
 - Better data structures and algorithms: better asymptotic behavior
 - Optimized code: smaller constants
- Techniques for speeding up a program
 - · Coax the compiler
 - Exploit capabilities of the hardware
 - Capitalize on knowledge of program execution



Course Wrap Up

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The Rest of the Semester



- Final Assignment Due: Sunday Jan 10
- Deans Date: Tuesday Jan 12
 - Cannot submit final assignment after 11:59PM
- Final Exam: Friday, Jan 22
 - 7:30 PM in room McCosh 10
 - · Exams from previous semesters are online at
 - http://www.cs.princeton.edu/courses/archive/fall09/cos217/exam2prep/
 - Covers entire course, with emphasis on second half of the term
 - Open book, open notes, open slides, etc. (just no computers!)
 - · No need to print/bring the IA-32 manuals
- Office hours during reading/exam period
 - · Daily, times TBA on course mailing list
- Review sessions
 - · During exam period, time TBA on course mailing list

Goals of COS 217



- Understand Modularity and Abstraction
 - Separation of interface from implementation
 - · Process/Processor, Virtual/Physical memory, et
- Understand boundary between code and co
 - · Machine architecture
 - Operating systems
 - Compilers
- Learn C and the Unix development tools
- · Improve your programming and debugging skills

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Relationship to Other Courses



- Machine architecture
 - Logic design (306) and computer architecture (471)
 - COS 217: assembly language and basic architecture
- Operating systems
 - Operating systems (318)
 - COS 217: virtual memory, system calls, and signals
- Compilers
 - Compiling techniques (320)
 - COS 217: compilation process, symbol tables, assembly and machine language
- Software systems
 - Numerous courses, independent work, etc.
 - COS 217: programming skills, UNIX tools, and ADTs

Lessons About Computer Science



- Modularity
 - Well-defined interfaces between components
 - Allows changing the implementation of one component without changing another
 - The key to managing complexity in large systems
- Resource sharing
 - Time sharing of the CPU by multiple processes
 - Sharing of the physical memory by multiple processes
- Indirection
 - Representing address space with virtual memory
 - Manipulating data via pointers (or addresses)

Lessons Continued



- Hierarchy
 - Memory: registers, cache, main memory, disk, tape, ...
 - Balancing the trade-off between fast/small and slow/big
- · Bits can mean anything
 - Code, addresses, characters, pixels, money, grades, ...
 - Arithmetic can be done through logic operations
 - The meaning of the bits depends entirely on how they are accessed, used, and manipulated

