Performance Improvement

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

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

Helping the Compiler Do Its Job

Optimizing Compilers

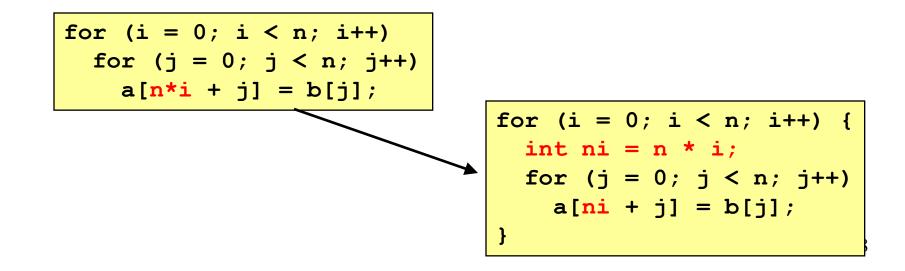
- Provide efficient mapping of program to machine
 - Register allocation
 - Code selection and ordering
 - Eliminating minor inefficiencies
- Don't (usually) improve asymptotic efficiency
 - Up to the programmer to select best overall algorithm
- Have difficulty overcoming "optimization blockers"
 -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

Limitations of Optimizing Compilers

- 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



Worrying About Side Effects

- 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
- <u>Is this transformation okay?</u>

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

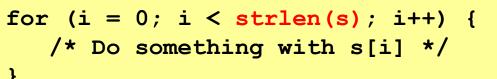
Not necessarily, if

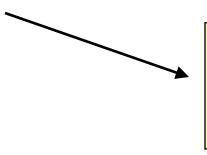
```
int counter = 0;
int f(int x) {
  return counter++;
}
```

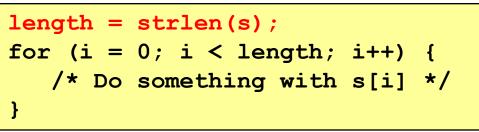
And this function may be defined in another file known only at link time!

Another Example on Side Effects

Is this optimization okay?







- 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

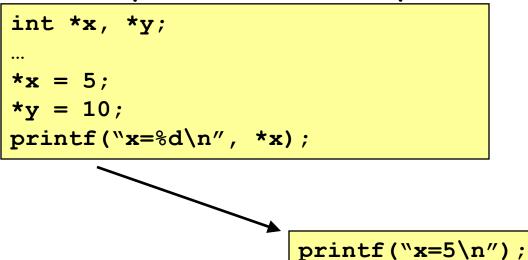
Memory Aliasing

- 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
- 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

xp, yp -

Another Aliasing Example

• Is this optimization okay?



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

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

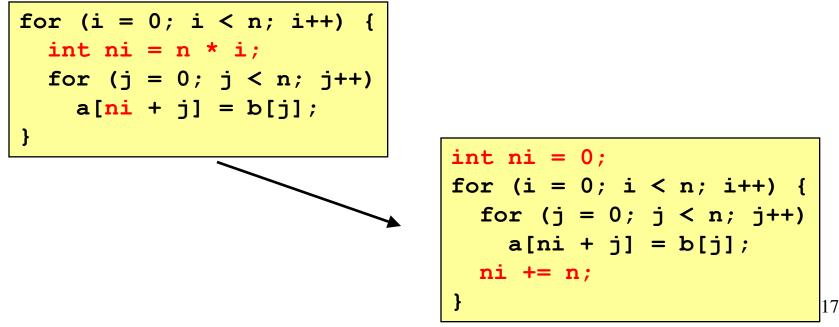
Exploiting the Hardware

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)

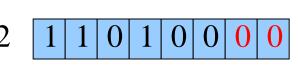
Addition Faster Than Multiplication

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



Bit Operations Faster Than Arithmetic

- Shift operations to multiple/divide by powers
 of 2
 - "x >> 3" is faster than "x/8" 53<<2
 - "x << 3" is faster than "x * 8"



0

0|0|

0

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()

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



Caching: Matrix Multiplication

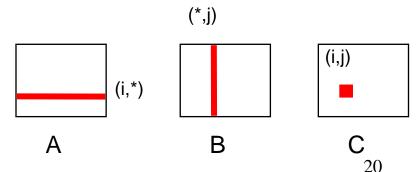
- Caches
 - Slower than registers, but faster than main memory
 - Both instruction caches and data caches
- Locality
 - Temporal locality: recently-referenced items are likely to be referenced in near future
 - Spatial locality: Items with nearby addresses tend to be referenced close together in time
- 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

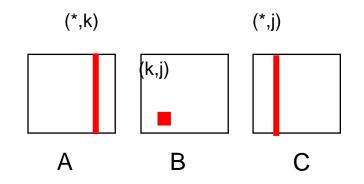
- Good spatial locality for A
- Poor spatial locality for B
- Good temporal locality for C



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];
    }
}</pre>
```

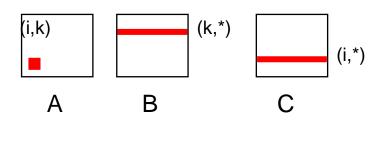
- Rather poor cache effects
 - Bad spatial locality for A
 - Good temporal locality for B
 - Bad spatial locality for C



Matrix Multiply: Cache Effects

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

- Good poor cache effects
 - Good temporal locality for A
 - Good spatial locality for B
 - Good 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

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

Avoiding Function Calls

- Function calls are expensive
 - Caller saves registers and pushes arguments on stack
 - Callee saves registers and pushes local variables on stack
 - Call and return disrupt the sequence flow of the code
- Function inlining:
 void g(void) {
 /* Some code */
 }
 void f(void) {
 ...
 g();
 ...
 }
 void f(void) {
 ...
 }

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 request 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

Goals of EE 209

- Understand boundary between code and computer
 - Machine architecture
 - Operating systems
 - Compilers
- Learn C and the Unix development tools
 - C is widely used for programming low-level systems
 - Unix has a rich development environment
 - Unix is open and well-specified, good for study & research
- Improve your programming skills
 - More experience in programming
 - Challenging and interesting programming assignments
 - Emphasis on modularity and debugging

Lessons Learned

- 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

Computer Networks (EE323)

- How Internet Works?
 - How it is designed, operated, and going to evolve?
 - How do I browse the Web, exchange emails?
 - How do hackers attack the Internet? How to defend? Secure/private network communication?
- Design of the Internet is very simple but solid
 - Smart end hosts + dumb networks
 - TCP/IP is the fundamental Internet protocol
 - Smart distributed algorithms
- Learn one of the smartest inventions!
 - With exciting assignments!!

Thank You!

- EE209 was a tough course
 - But hopefully you remember it for longer time
- Good luck with your final exam!
 - And your final assignment!