



# 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

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## 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

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## Limitations of Optimizing Compilers

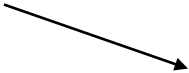
- **Fundamental constraint**
  - Compiler must not change program behavior
  - Ever, even under rare pathological inputs
- **Behavior that may be obvious to the programmer can be obfuscated by languages and coding styles**
  - Data ranges more limited than variable types suggest
  - Array elements remain 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

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## 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];  
}
```

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## 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
- Is this transformation okay?

```
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!

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## Another Example on Side Effects



- Is this optimization okay?

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

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

- 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

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## 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;  
}
```

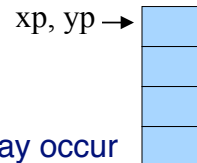
- Not necessarily, what if *xp* and *yp* are equal?
  - First version: result is 4 times \**xp*
  - Second version: result is 3 times \**xp*

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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



- 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

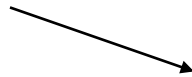
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## Another Aliasing Example



- 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

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## 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)

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## Addition Faster Than Multiplication



- Adding instead of multiplying
  - Addition is faster than multiplication
- Recognize sequences of 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;  
}
```

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## Bit Operations Faster Than Arithmetic



- Shift operations to multiple/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 1 0 1

& 15 0 0 0 0 1 1 1 1

5 0 0 0 0 0 1 0 1

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## Caching: Matrix Multiplication



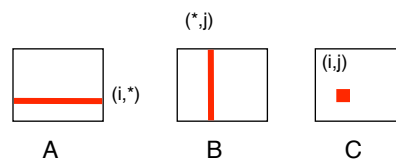
- **Caches**
  - Slower than registers, but faster than main memory
  - Both instruction caches and data caches
- **Locality**
  - Temporal locality: recently-referenced items are likely to be referenced in near future
  - Spatial locality: Items with nearby addresses tend to be referenced close together 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<sub>9</sub>

## 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];  
  }  
}
```

- **Reasonable cache effects**
  - Good spatial locality for A
  - Poor spatial locality for B
  - Good temporal locality for C



## Matrix Multiply: Cache Effects

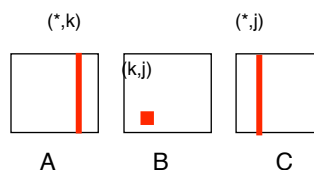


```

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



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## Matrix Multiply: Cache Effects

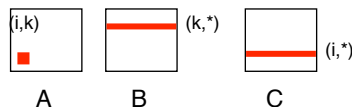


```

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];
  }
}

```

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



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## Parallelism: Loop Unrolling



- What limits the performance?

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

- 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];
```

- 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];
```

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## Program Execution

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## 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();  
    ...  
}
```

Some compilers support  
“inline” keyword directive.

```
void f(void) {  
    ...  
    /* Some code */  
    ...  
}
```

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## 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>

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## Consider The Easy Way Out



- Hardware might be cheaper
  - Developers are expensive
  - Hardware keeps dropping in price
  - Fixed inefficiency may be tolerable
- Example
  - High-performance Web server
  - Post-connection info maintained for 120 seconds
  - At 8000 reqs/sec, almost 1M post-connection records!
  - Horrible? 128 bytes/record = 128MB of kernel memory
  - DRAM list price: \$30/GB
  - Total cost of post-connection memory: \$4

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## Understand Defaults



- Sometimes, limits exist in OS/shell

- Set to “reasonable” default values
- `$ ulimit -a`
- **core file size** (blocks, -c) 0
- **data seg size** (kbytes, -d) unlimited
- **scheduling priority** (-e) 0
- **file size** (blocks, -f) unlimited
- **pending signals** (-i) 65536
- **max locked memory** (kbytes, -l) 64
- **max memory size** (kbytes, -m) unlimited
- **open files** (-n) 8192
- **pipe size** (512 bytes, -p) 8
- **POSIX message queues** (bytes, -q) 819200
- **real-time priority** (-r) 0
- **stack size** (kbytes, -s) 10240
- **cpu time** (seconds, -t) unlimited
- **max user processes** (-u) 1024
- **virtual memory** (kbytes, -v) unlimited
- **file locks** (-x) unlimited

- Sometimes you need to be unreasonable

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## Understand “Hidden” Limits



- Company was using system w/o database
- Use geo-targeting system for demographics
  - Map IP address to zip code
  - Lots of databases (income, etc) by zip code
  - 6 digit zip = 100K possible, but only 50K really used
- Symptoms
  - Performance looked fine on small tests (thousands of lookups/sec)
  - On deployed system, entire machine performance dropped
  - All applications handled only 100's reqs/sec
- Created one file per used zip code
  - Each file relatively small
  - System configured to cache < 50K files

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## 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

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