CSC 252: Computer Organization Spring 2021: Lecture 15

Instructor: Yuhao Zhu

Department of Computer Science
University of Rochester

- Programming assignment 3 is due today
 - Details: https://www.cs.rochester.edu/courses/252/spring2021/labs/assignment3.html
 - Due on **March 23**, 11:59 PM
- Lab 2 grades are out.

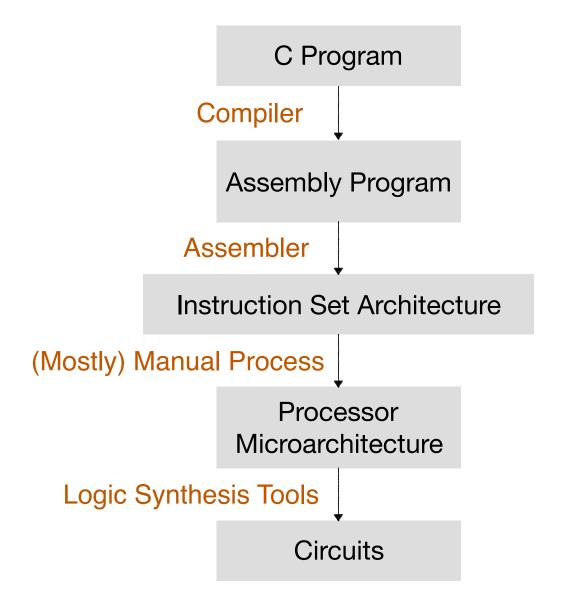
7	8	•	10	11	12	13
14	15	16	17	18	19	20
21	22	Today Due	24	Mid-term	26	27

- Mid-term exam: this Thursday, 12:30 PM 1:45 PM; online.
- Past exam & Problem set: https://www.cs.rochester.edu/courses/252/spring2021/handouts.html
- Exam will be electronic using Gradescope, but we will send you an PDF version so that you can work offline in case
 - 1) you don't have stable Internet access at the exam time or
 - 2) the time doesn't work for you
 - Write down the answers on a scratch paper, take pictures, and send us the pictures
 - Send me an email explaining why you need to work offline before the exam data/time.

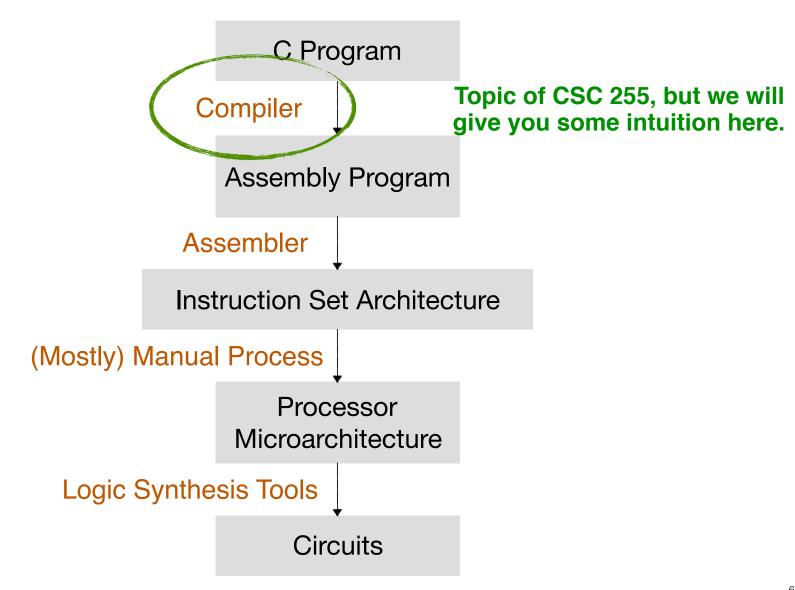
- Exams are designed to test your ability to apply what you have learned and not your memory (though a good memory could help).
- Open book test: any sort of paper-based product, e.g., book, notes, magazine, old tests.
- Nothing electronic (including laptop, cell phone, calculator, etc) other than the computer you use to take the exam.
- You can use your electronic copy of the textbook, but you must not search in the textbook.
- **Nothing biological**, including your roommate, husband, wife, your hamster, another professor, etc.
- "I don't know" gets 15% partial credit. Must erase everything else.

- Mock exam now. Won't be graded.
- Go to https://www.gradescope.com/courses/256242/assignments/1116540.
- Log in using your NetID.

So far in 252...



So far in 252...



Code Optimization Overview

- Three entities can optimize the program: programmer, compiler, and hardware
- The best thing to speed up a program is to pick a good algorithm. Compilers/hardware can't do that in general.
 - Quicksort: O(n log n) = K * n * log(n)
 - Bubblesort: O(n^2) = K * n^2
- Algorithm choice decides overall complexity (big O), compiler/ hardware decides the constant factor in the big O notation
- Compiler and hardware implementations decide the K.
- Programmers can write code that makes it easier to compiler and hardware to improve performance.

Optimizing Code Transformation

- Hardware/Microarchitecture Independent Optimizations
 - Code motion/precomputation
 - Strength reduction
 - Sharing of common subexpressions
- Optimization Blockers
 - Procedure calls
 - Memory aliasing
- Exploit Hardware Microarchitecture

Generally Useful Optimizations

 Optimizations that you or the compiler should do regardless of processor

Code Motion

- Reduce frequency with which computation performed
 - If it will always produce same result
 - Especially moving code out of loop

```
void set_row(double *a, double *b,
    long i, long n)
{
    long j;
    for (j = 0; j < n; j++)
        a[n*i+j] = b[j];
}</pre>
```

```
long j;
int ni = n*i;
for (j = 0; j < n; j++)
    a[ni+j] = b[j];</pre>
```

Compiler-Generated Code Motion (-O1)

```
void set_row(double *a, double *b,
    long i, long n)
{
    long j;
    for (j = 0; j < n; j++)
        a[n*i+j] = b[j];
}</pre>
```

```
long j;
int ni = n*i;
for (j = 0; j < n; j++)
    a[ni+j] = b[j];</pre>
```





```
set row:
          testq
                    %rcx, %rcx
                                                    # Test n
                                                    # If 0, goto done
          jle
                     . L1
                                                    # ni = n*i
          imulq
                    %rcx, %rdx
                    (\$rdi,\$rdx,8), \$rdx # rowp = A + ni*8
          leag
                    $0, %eax
                                                    # j = 0
          movl
.L3:
                                                    # loop:
                    (%rsi,%rax,8), %xmm0
                                                    # t = b[j]
          movsd
                    %xmm0, (%rdx, %rax, 8)
                                                    \# M[A+ni*8 + j*8] = t
          movsd
          addq
                    $1, %rax
                                                    # j++
                    %rcx, %rax
                                                    # j:n
          cmpq
                                                    # if !=, goto loop
                     . L3
          jne
                                                    # done:
.L1:
          rep ; ret
```

Reduction in Strength

- Replace costly operation with simpler one
- Shift, add instead of multiply or divide
 - 16*x --> x << 4
 - Depends on cost of multiply or divide instruction
 - On Intel Nehalem, integer multiply requires 3 CPU cycles. Division takes even more cycles. Shift can generally be done in 1 cycle.
- Use the lea instruction

Reduction in Strength

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```
• 16*x --> x << 4
```

- Depends on cost of multiply or divide instruction
- On Intel Nehalem, integer multiply requires 3 CPU cycles. Division takes even more cycles. Shift can generally be done in 1 cycle.
- Use the lea instruction

```
long m12(long x)
{
  return x*12;
}
```

```
leaq (%rdi,%rdi,2), %rax # t <- x+x*2
salq $2, %rax # return t<<2
```

Common Subexpression Elimination

- Reuse portions of expressions
- GCC will do this with –O1

```
3 multiplications: i^*n, (i-1)^*n, (i+1)^*n
```

```
/* Sum neighbors of i,j */
up = val[(i-1)*n + j ];
down = val[(i+1)*n + j ];
left = val[i*n + j-1];
right = val[i*n + j+1];
sum = up + down + left + right;
```

```
leaq
      1(%rsi), %rax
                   # i+1
      -1(%rsi), %r8 # i-1
leag
imulq
      %rcx, %rsi # i*n
      %rcx, %rax # (i+1)*n
imulq
imulq %rcx, %r8 # (i-1)*n
      %rdx, %rsi # i*n+j
addq
                 # (i+1)*n+j
addq
      %rdx, %rax
addq
      %rdx, %r8
                   # (i-1)*n+j
```

1 multiplication: i*n

```
long inj = i*n + j;
up =    val[inj - n];
down = val[inj + n];
left = val[inj - 1];
right = val[inj + 1];
sum = up + down + left + right;
```

```
imulq %rcx, %rsi # i*n
addq %rdx, %rsi # i*n+j
movq %rsi, %rax # i*n+j
subq %rcx, %rax # i*n+j-n
leaq (%rsi,%rcx), %rcx # i*n+j+n
```

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 - Memory aliasing
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Procedure to Convert String to Lower Case

```
void lower(char *s)
{
    size_t i;
    for (i = 0; i < strlen(s); i++)
        if (s[i] >= 'A' && s[i] <= 'Z')
        s[i] -= ('A' - 'a');
}</pre>
```

Calling Strlen

```
size_t strlen(const char *s)
{
    size_t length = 0;
    while (*s != '\0') {
        s++;
        length++;
    }
    return length;
}
```

Strlen performance

- Has to scan the entire length of a string, looking for null character.
- O(N) complexity

Overall performance

- N calls to strlen
- Overall O(N²) performance

Improving Performance

- Move call to strlen outside of loop
- Since result does not change from one iteration to another
- Form of code motion

```
void lower(char *s)
{
    size_t i;
    size_t len = strlen(s);
    for (i = 0; i < len; i++)
        if (s[i] >= 'A' && s[i] <= 'Z')
        s[i] -= ('A' - 'a');
}</pre>
```

```
void lower(char *s)
  size t i;
  for (i = 0; i < strlen(s); i++)
    if (s[i] >= 'A' \&\& s[i] <= 'Z')
      s[i] -= ('A' - 'a');
size t total lencount = 0;
size t strlen(const char *s)
    size t length = 0;
    while (*s != '\0') {
       s++; length++;
    total lencount += length;
    return length;
```

Why couldn't compiler move strlen out of loop?

- Procedure may have side effects, e.g., alters global state each time called
- Function may not return same value for given arguments

- Most compilers treat procedure call as a black box
 - Assume the worst case, weak optimizations near them
 - There are interprocedural optimizations (IPO), but they are expensive
 - Sometimes the compiler doesn't have access to source code of other functions because they are object files in a library. Link-time optimizations (LTO) comes into play, but are expensive as well.

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Remedies:

- Use of inline functions
- Do your own code motion

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Remedies:

- Use of inline functions
- Do your own code motion

```
inline void swap(int *m, int *n) {
  int tmp = *m;
  *m = *n;
  *n = tmp;
}

void foo () {
  swap(&x, &y);
}
```

```
/* Sum rows of n X n matrix a
   and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
   for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
   }
}</pre>
```

Value of a:

```
init: [x, x, x]
```

```
double a[9] =
{ 0, 1, 2,
    4, 8, 16,
    32, 64, 128};
```

```
/* Sum rows of n X n matrix a
    and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

Value of a:

double a[9] = { 0, 1, 2, 4, 8, 16, 32, 64, 128};

```
init: [x, x, x]
```

$$i = 0: [3, x, x]$$

```
/* Sum rows of n X n matrix a
    and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

Value of a:

double a[9] = { 0, 1, 2, 4, 8, 16, 32, 64, 128};

```
init: [x, x, x]
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        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
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Value of a:

double a[9] = { 0, 1, 2, 4, 8, 16, 32, 64, 128};

```
init: [x, x, x]
```

$$i = 0: [3, x, x]$$

$$i = 1: [3, 28, x]$$

A Potential Optimization

```
/* Sum rows of n X n matrix a
    and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        Every iteration updates
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
        Memory accesses are
        slow, SO...</pre>
```

A Potential Optimization



```
double val = 0;
for (j = 0; j < n; j++)
   val += a[i*n + j];
b[i] = val;</pre>
```

Every iteration updates val, which could stay in register. Update memory only once.

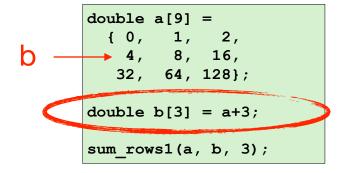
A Potential Optimization

```
/* Sum rows of n X n matrix a
   and store in vector b */
void sum rows1(double *a, double *b, long n) {
   long i, j;
   for (i = 0; i < n; i++) {
                                          Every iteration updates
         b[i] = 0:
                                          memory location b[i].
         for (j = 0; j < n; j++)
             b[i] += a[i*n + j];
                                          Memory accesses are
                                          slow, so...
                                    Every iteration updates val,
        double val = 0;
        for (j = 0; j < n; j++)
                                    which could stay in register.
          val += a[i*n + j];
        b[i] = val;
                                    Update memory only once.
```

Why can't a compiler perform this optimization?

```
/* Sum rows of n X n matrix a
    and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

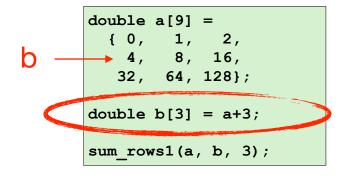
Value of a:



```
init: [4, 8, 16]
```

```
/* Sum rows of n X n matrix a
   and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

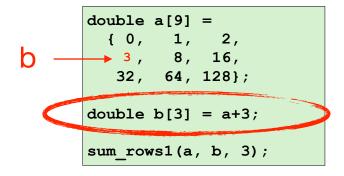
Value of a:



```
init: [4, 8, 16]
i = 0: [3, 8, 16]
```

```
/* Sum rows of n X n matrix a
    and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

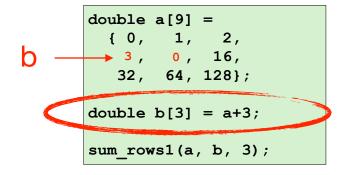
Value of a:



```
init: [4, 8, 16]
i = 0: [3, 8, 16]
```

```
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void sum_rows1(double *a, double *b, long n) {
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        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

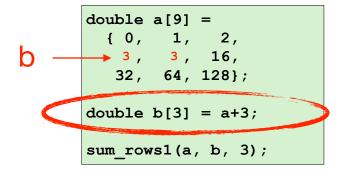
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i = 0: [3, 8, 16]
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        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

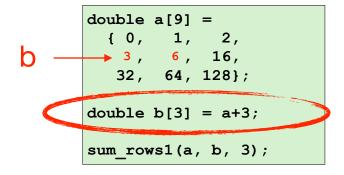
Value of a:



```
init: [4, 8, 16]
i = 0: [3, 8, 16]
```

```
/* Sum rows of n X n matrix a
    and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

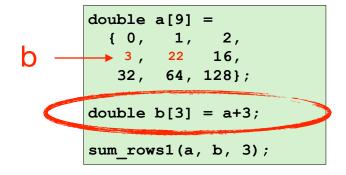
Value of a:



```
init: [4, 8, 16]
i = 0: [3, 8, 16]
```

```
/* Sum rows of n X n matrix a
   and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

Value of a:

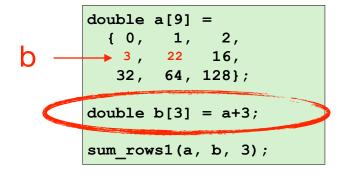


```
init: [4, 8, 16]
i = 0: [3, 8, 16]
```

Memory Aliasing

```
/* Sum rows of n X n matrix a
   and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

Value of a:



Value of b:

```
init: [4, 8, 16]

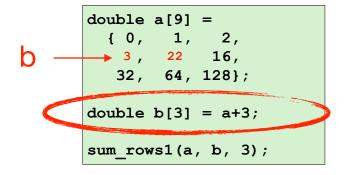
i = 0: [3, 8, 16]

i = 1: [3, 22, 16]
```

Memory Aliasing

```
/* Sum rows of n X n matrix a
   and store in vector b */
void sum_rows1(double *a, double *b, long n) {
    long i, j;
    for (i = 0; i < n; i++) {
        b[i] = 0;
        for (j = 0; j < n; j++)
            b[i] += a[i*n + j];
    }
}</pre>
```

Value of a:



Value of b:

```
init: [4, 8, 16]

i = 0: [3, 8, 16]

i = 1: [3, 22, 16]

i = 2: [3, 22, 224]
```

Optimization Blocker: Memory Aliasing

- Aliasing
 - Two different memory references (array elements or pointers) specify the same memory location
 - Easy to have in C
 - Since C allows address/pointer arithmetic
 - Direct access to storage structures
 - Get in habit of introducing local variables
 - Accumulating within loops
 - Your way of telling compiler not to check for aliasing

Today: Optimizing Code Transformation

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- Exploit Hardware Microarchitecture

Exploiting Instruction-Level Parallelism (ILP)

- Hardware can execute multiple instructions in parallel
 - Pipeline is a classic technique. Multiple instructions are being executed at the same time
- Performance limited by control/data dependencies
- Simple transformations can yield dramatic performance improvement
 - Compilers often cannot make these transformations
 - Lack of associativity and distributivity in floating-point arithmetic

Baseline Code

```
for (i = 0; i < length; i++) {
  t = t * d[i];
  *dest = t;
}</pre>
```

Loop Unrolling (2x1)

```
long limit = length-1;
long i;
/* Combine 2 elements at a time */
for (i = 0; i < limit; i+=2) {
    x = (x * d[i]) * d[i+1];
}

/* Finish any remaining elements */
for (; i < length; i++) {
    x = x * d[i];
}
*dest = x;</pre>
```

- Perform 2x more useful work per iteration
- Reduce loop overhead (comp, jmp, index dec, etc.)
- What's the trade-off here?

Loop Unrolling with Separate Accumulators

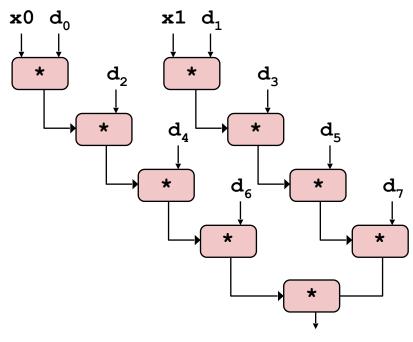
```
long limit = length-1;
long i;
/* Combine 2 elements at a time */
for (i = 0; i < limit; i+=2) {
    x0 = x0 * d[i];
    x1 = x1 * d[i+1];
}

/* Finish any remaining elements */
for (; i < length; i++) {
    x0 = x0 * d[i];
}
*dest = x0 * x1;</pre>
```

Data-Flow Graph (DFG)

```
x0 = x0 * d[i];

x1 = x1 * d[i+1];
```



- What changed:
 - Two independent "streams" of operations
 - Reduce data dependency
- What was the DFG like using just one accumulator?

Code Optimization Summary

- From a programmer's perspective:
 - What you know: the functionality/intention of your code; the inputs to the program; all the code in the program
 - What you might not know: the hardware details.
- From a compiler's perspective:
 - What you know: all the code in the program; (maybe) the hardware details.
 - What you might not know: the inputs to the program; the intention of the code
- From the hardware's perspective:
 - What you know: the hardware details; some part of the code
 - What you might not know: the inputs to the program; the intention of the code
- The different perspectives indicate that different entities have different responsibilities, limitations, and advantages in optimizing the code

```
float foo(int x, int y)
{
    return pow(x, y) * 100 / log(x) * sqrt(y);
}
```

• As a programmer, if you know what x and y will be, say 5, you could direct return the results 23769.8 without having to the computation

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- Except...Profile-guided optimizations:

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- Except...Profile-guided optimizations:
 - Run the code multiple times using some sample inputs, and observe the values of x and y (statistically).

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- Compiler would have no idea
- Except...Profile-guided optimizations:
 - Run the code multiple times using some sample inputs, and observe the values of x and y (statistically).
 - If let's say 99% of the time, x = 2 and y = 5, what could the compiler do then?

```
float foo(int x, int y)
{
    return pow(x, y) * 100 / log(x) * sqrt(y);
}
```

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- Except...Profile-guided optimizations:
 - Run the code multiple times using some sample inputs, and observe the values of x and y (statistically).
 - If let's say 99% of the time, x = 2 and y = 5, what could the compiler do then?

```
float foo(int x, int y)
{
    return pow(x, y) * 100 / log(x) * sqrt(y);
}
```

```
float foo(int x, int y)
{
   if (x == 2 && y == 5) return 23769.8;
   else return pow(x, y) * 100 / log(x) * sqrt(y);
}
```