CS 33

Architecture and Optimization (1)

Simplistic View of Processor

```
while (true) {
  instruction = mem[eip];
  execute(instruction);
}
```

Some Details ...

```
void execute(instruction_t instruction) {
  decode(instruction, &opcode, &operands);
  fetch(operands, &in_operands);
  perform(opcode, in_operands, &out_operands);
  store(out_operands);
}
```

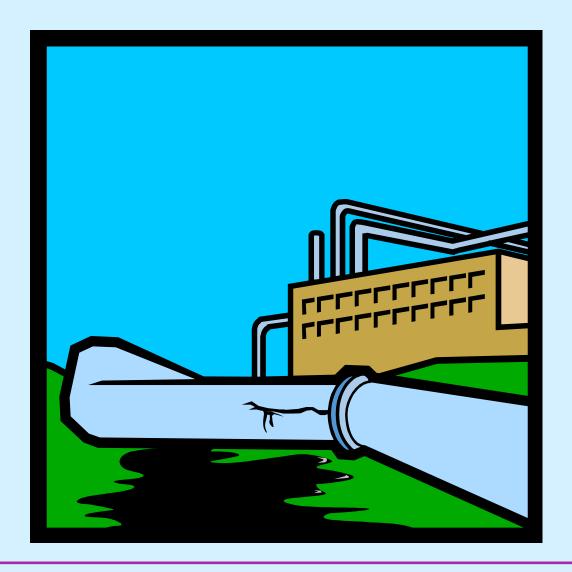
Pipelines

Decode	Fetch	Perform	Store	Decode	Fetch	Perform	Store
Decode	Fetch	Perform	Store				
	Decode	Fetch	Perform	Store			
		Decode	Fetch	Perform	Store		
			Decode	Fetch	Perform	Store	
				Decode	Fetch	Perform	Store

Analysis

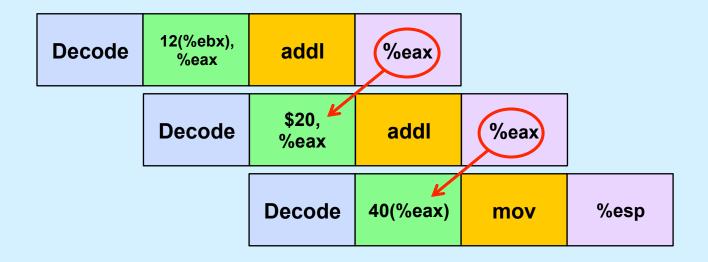
- Not pipelined
 - each instruction takes, say, 320 nanoseconds
 - » 320 ns latency
 - 3.125 billion instructions/second (GIPS)
- Pipelined
 - each instruction still takes 320 ns
 - » latency still 320 ns
 - an instruction completes every 80 ns
 - » 12.5 GIPS throughput

Hazards ...

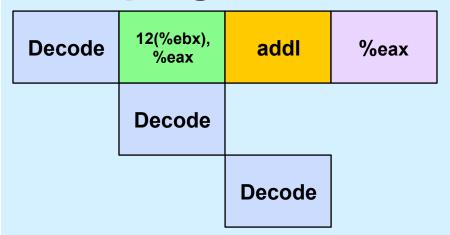


Data Hazards

```
addl 12(%ebx), %eax
addl $20, %eax
movl 40(%eax), %esp
```



Coping





40(%eax) mov

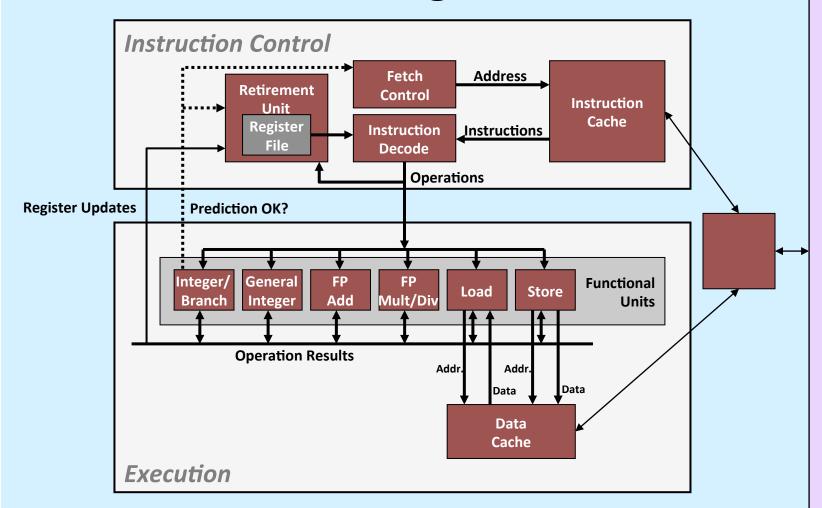
Control Hazards

```
movl $0, %ecx
.L2:
  movl %edx, %eax
  andl $1, %eax
  addl %eax, %ecx
  shrl $1, %edx
  jne .L2 # what goes in the pipeline?
  movl %ecx, %eax
  ...
```

Coping: Guess ...

- Branch prediction
 - assume, for example, that conditional branches are always taken
 - but don't do anything to registers or memory until you know for sure

Modern CPU Design



Memory

Performance Realities

There's more to performance than asymptotic complexity

- Constant factors matter too!
 - easily see 10:1 performance range depending on how code is written
 - must optimize at multiple levels:
 - » algorithm, data representations, procedures, and loops
- Must understand system to optimize performance
 - how programs are compiled and executed
 - how to measure program performance and identify bottlenecks
 - how to improve performance without destroying code modularity and generality

Optimizing Compilers

- Provide efficient mapping of program to machine
 - register allocation
 - code selection and ordering (scheduling)
 - dead code elimination
 - eliminating minor inefficiencies
- Don't (usually) improve asymptotic efficiency
 - up to programmer to select best overall algorithm
 - big-O savings are (often) more important than constant factors
 - » but constant factors also matter
- Have difficulty overcoming "optimization blockers"
 - potential memory aliasing
 - potential procedure side-effects

Limitations of Optimizing Compilers

- Operate under fundamental constraint
 - must not cause any change in program behavior
 - often prevents it from making optimizations that would only affect behavior under pathological conditions
- Behavior that may be obvious to the programmer can be obfuscated by languages and coding styles
 - e.g., data ranges may be more limited than variable types suggest
- Most analysis is performed only within procedures
 - whole-program analysis is too expensive in most cases
- Most analysis is based only on static information
 - compiler has difficulty anticipating run-time inputs
- When in doubt, the compiler must be conservative

Generally Useful Optimizations

- Optimizations that you or the compiler should do regardless of processor / compiler
- Code Motion
 - reduce frequency with which computation performed
 - » if it will always produce same result
 - » especially moving code out of loop

```
void set_row(long *a, long *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

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

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

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

Reduction in Strength

- Replace costly operation with simpler one
- Shift, add instead of multiply or divide

```
16*x --> x << 4
```

- utility is machine-dependent
- depends on cost of multiply or divide instruction
 on Intel Nehalem, integer multiply requires 3 CPU cycles
- Recognize sequence of products

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

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

Share Common Subexpressions

- Reuse portions of expressions
- Compilers often not very sophisticated in exploiting arithmetic properties

```
/* 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;
```

```
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;
```

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

```
1 multiplication: i*n
```

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

```
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
```

Quiz 1

The fastest means (on the Intel Nehalem) for evaluating

$$n*n + 2*n + 1$$

requires exactly:

- a) 2 multiplies and 2 additions
- b) one multiply and two additions
- c) one multiply and one addition
- d) three additions

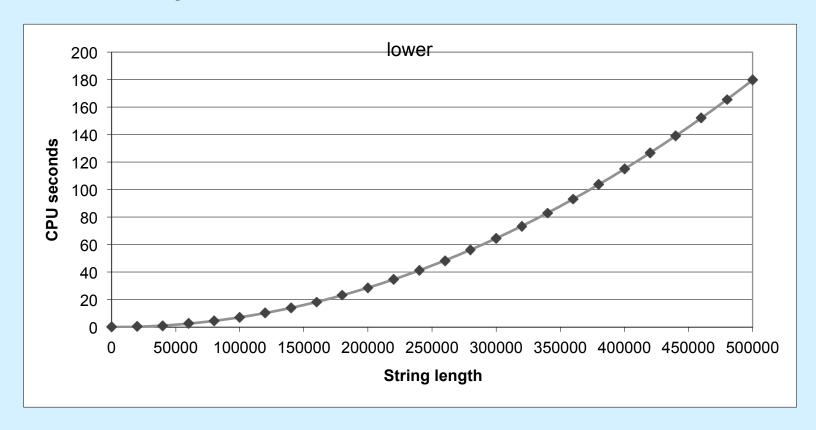
Optimization Blocker #1: Procedure Calls

Procedure to convert string to lower case

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

Lower Case Conversion Performance

- Time quadruples when string length doubles
- Quadratic performance



Convert Loop To Goto Form

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

strlen executed every iteration

Calling Strlen

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

- strlen performance
 - only way to determine length of string is to scan its entire length, looking for null character
- Overall performance, string of length N
 - N calls to strlen
 - overall O(N²) performance

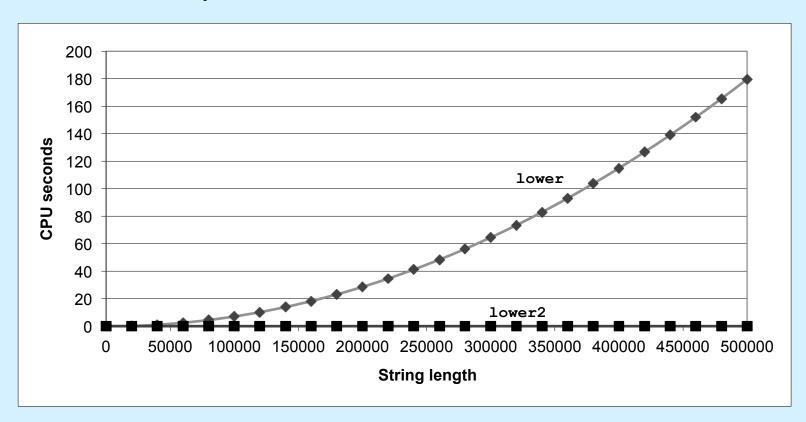
Improving Performance

```
void lower2(char *s) {
  int i;
  int len = strlen(s);
  for (i = 0; i < len; i++)
    if (s[i] >= 'A' && s[i] <= 'Z')
      s[i] -= ('A' - 'a');
}</pre>
```

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

Lower-Case Conversion Performance

- Time doubles when string-length doubles
 - linear performance of lower2



Optimization Blocker: Procedure Calls

- Why couldn't compiler move strlen out of inner loop?
 - procedure may have side effects
 - » alters global state each time called
 - function may not return same value for given arguments
 - » depends on other parts of global state
 - » procedure lower could interact with strlen

Warning:

- compiler treats procedure call as a black box
- weak optimizations near them
- Remedies:
 - use of inline functions
 - » gcc does this with –O2
 - do your own code motion

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

Memory Matters

```
/* Sum rows of n X n matrix a
   and store result in vector b */
void sum_rows1(long *a, long *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>
```

```
# sum_rows1 inner loop
.L3:
    movq (%rdi), %rcx  # rcx = *aptr
    addq %rcx, (%rsi,%rax,8) # b[i] += rcx
    addq $8, %rdi  # aptr++
    cmpq %r8, %rdi
    jne .L3
```

- Code updates b[i] on every iteration
- Why couldn't compiler optimize this away?

Memory Aliasing

```
/* Sum rows of n X n matrix a
   and store result in vector b */
void sum_rows1(int *a, int *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>
```

```
int A[9] =
  { 0,   1,   2,
     4,   8,  16,
   32,  64,  128};
int *B = &A[3];
sum_rows1(A, B, 3);
```

Value of B:

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

$$i = 0: [3, 8, 16]$$

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

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

- Code updates b[i] on every iteration
- Must consider possibility that these updates will affect program behavior

Removing Aliasing

```
/* Sum rows of n X n matrix a
   and store result in vector b */
void sum_rows2(int *a, int *b, int n) {
   long i, j;
   for (i = 0; i < n; i++) {
      int val = 0;
      for (j = 0; j < n; j++)
        val += a[i*n + j];
      b[i] = val;
}</pre>
```

```
# sum_rows2 inner loop
.L4:
   addq (%rdi), %rax
   addq $8, %rdi
   cmpq %rcx, %rdi
   jne .L4
```

No need to store intermediate results

Optimization Blocker: Memory Aliasing

Aliasing

- two different memory references specify single location
- easy to have happen in C
 - » since allowed to do address 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

C99 to the Rescue

New attribute

- restrict
 - » applied to a pointer, tells the compiler that the object pointed to will be accessed only via this pointer
 - » compiler thus doesn't have to worry about aliasing
 - » but the programmer does ...
 - » syntax

```
int *restrict pointer;
```

Memory Matters, Fixed

```
/* Sum rows of n X n matrix a
    and store result in vector b */
void sum_rows3(long *restrict a, long *restrict 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>
```

```
# sum_rows1 inner loop
.L3:
   addq (%rdi), %rax
   addq $8, %rdi
   cmpq %rcx, %rdi
   jne .L3
```

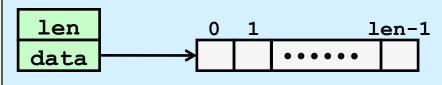
Code doesn't update b[i] on every iteration

Exploiting Instruction-Level Parallelism

- Need general understanding of modern processor design
 - hardware can execute multiple instructions in parallel
- Performance limited by data dependencies
- Simple transformations can have dramatic performance improvement
 - compilers often cannot make these transformations
 - lack of associativity and distributivity in floatingpoint arithmetic

Benchmark Example: Datatype for Vectors

```
/* data structure for vectors */
typedef struct{
   int len;
   data_t *data;
} vec_t, *vec_ptr_t;
```



```
/* retrieve vector element and store at val */
int get_vec_element(vec_ptr_t v, int idx, data_t *val){
    if (idx < 0 || idx >= v->len)
        return 0;
    *val = v->data[idx];
    return 1;
}

/* return length of vector */
int vec_length(vec_ptr_t v) {
    return v->len;
}
```

Benchmark Computation

```
void combine1(vec_ptr_t v, data_t *dest) {
    long int i;
    *dest = IDENT;
    for (i = 0; i < vec_length(v); i++) {
        data_t val;
        get_vec_element(v, i, &val);
        *dest = *dest OP val;
    }
}</pre>
```

Compute sum or product of vector elements

- Data Types
 - use different declarations for data_t

» int

» float

» double

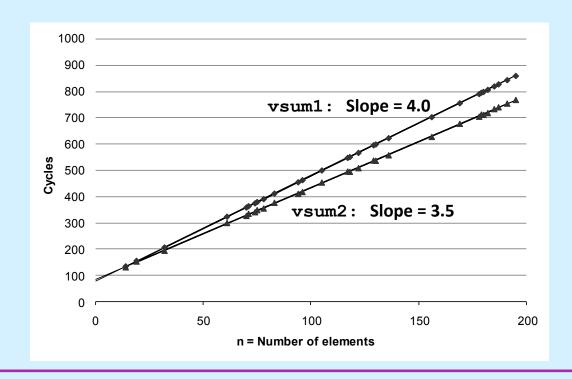
- Operations
 - use different definitions of
 op and IDENT

 \gg +, 0

» *, 1

Cycles Per Element (CPE)

- Convenient way to express performance of program that operates on vectors or lists
- Length = n
- T = CPE*n + Overhead
 - CPE is slope of line



Benchmark Performance

```
void combine1(vec_ptr_t v, data_t *dest) {
    long int i;
    *dest = IDENT;
    for (i = 0; i < vec_length(v); i++) {
        data_t val;
        get_vec_element(v, i, &val);
        *dest = *dest OP val;
    }
}</pre>
```

Compute sum or product of vector elements

Method	Integer		Double FP		
Operation	Add	Mult	Add	Mult	
Combine1 unoptimized	29.0	29.2	27.4	27.9	
Combine1 -O1	12.0	12.0	12.0	13.0	

Move vec_length

```
void combine2(vec_ptr_t v, data_t *dest) {
    long int i;
    long int length = vec_length(v);
    *dest = IDENT;
    for (i = 0; i < length; i++) {
        data_t val;
        get_vec_element(v, i, &val);
        *dest = *dest OP val;
    }
}</pre>
```

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine1 unoptimized	29.0	29.2	27.4	27.9
Combine1 -O1	12.0	12.0	12.0	13.0
Combine2	8.03	8.09	10.09	12.08

Eliminate Procedure Calls

```
void combine3(vec_ptr_t v, data_t *dest) {
    long int i;
    long int length = vec_length(v);
    data_t *data = get_vec_start(v);
    *dest = IDENT;
    for (i = 0; i < length; i++) {
        *dest = *dest OP data[i];
    }
}</pre>
```

```
data_t *get_vec_start(
    vec_ptr v) {
    return v->data;
}
```

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine2	8.03	8.09	10.09	12.08
Combine3	6.01	8.01	10.01	12.02

Eliminate Unneeded Memory References

```
void combine4(vec_ptr_t v, data_t *dest) {
  int i;
  int length = vec_length(v);
  data_t *d = get_vec_start(v);
  data_t t = IDENT;
  for (i = 0; i < length; i++)
    t = t OP d[i];
  *dest = t;
}</pre>
```

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine1 -O1	12.0	12.0	12.0	13.0
Combine4	2.0	3.0	3.0	5.0

Quiz 2

Combine4 is pretty fast; we've done all the "obvious" optimizations. How much faster will we be able to make it? (Hint: it involves taking advantage of pipelining and multiple functional units on the chip.)

- a) 1× (it's already as fast as possible)
- b) $2 \times -4 \times$
- c) $16 \times -64 \times$