

Code Optimization II : Machine-dependent Optimizations

COMP400727: Introduction to Computer Systems

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Today

- **Machine-Dependent Optimizations**
- Instruction-Level Parallelism
- Branch Predictions

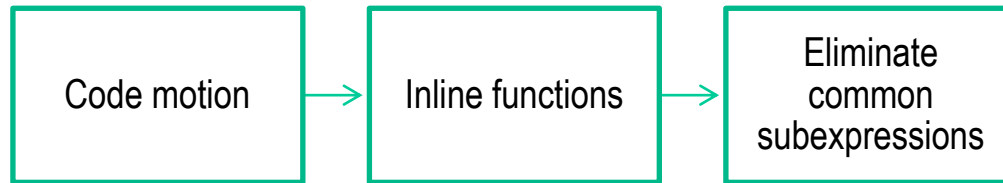
Multiple Levels of Optimizations

Human-level optimization



Preprocessing

Machine-independent optimization



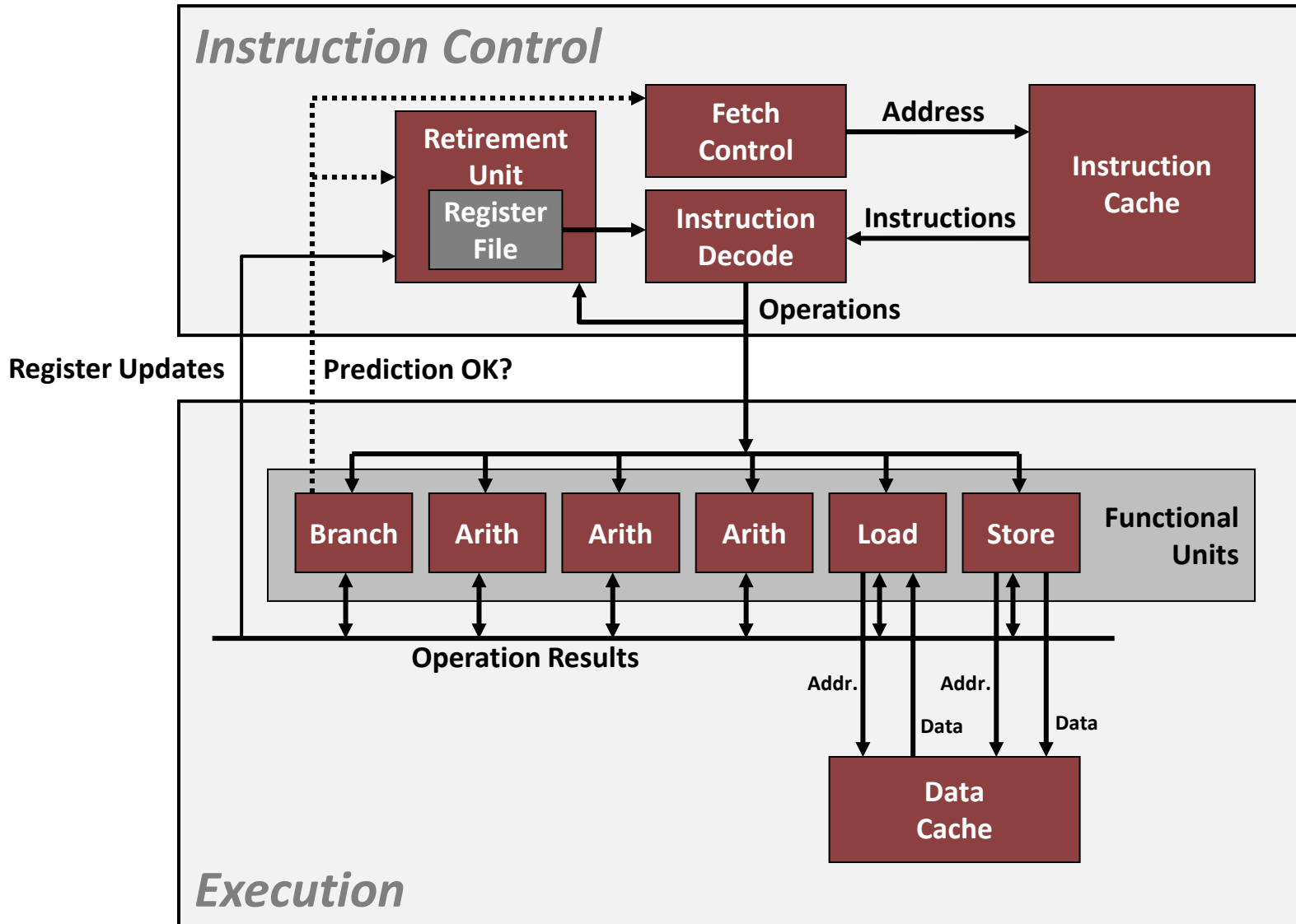
Machine-dependent optimization



Compilation

Assembling

Modern CPU Design



Today

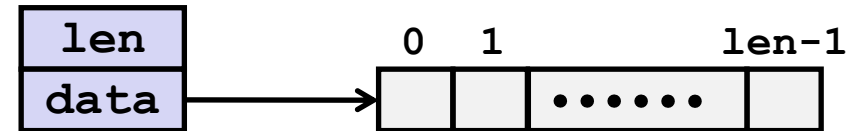
- Machine-Dependent Optimizations
- **Instruction-Level Parallelism**
- Branch Predictions

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 cause big speedups**
 - Compilers often cannot make these transformations
 - Lack of associativity and distributivity in floating-point arithmetic

Benchmark Example: Data Type for Vectors

```
/* data structure for vectors */
typedef struct{
    size_t len;
    data_t *data;
} vec;
```



■ Data Types

- Use different declarations for data_t
- int
- long
- float
- double

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

Benchmark Computation

```
void combine1(vec_ptr 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;
    }
}
```

Compute sum or
product of vector
elements

■ Data Types

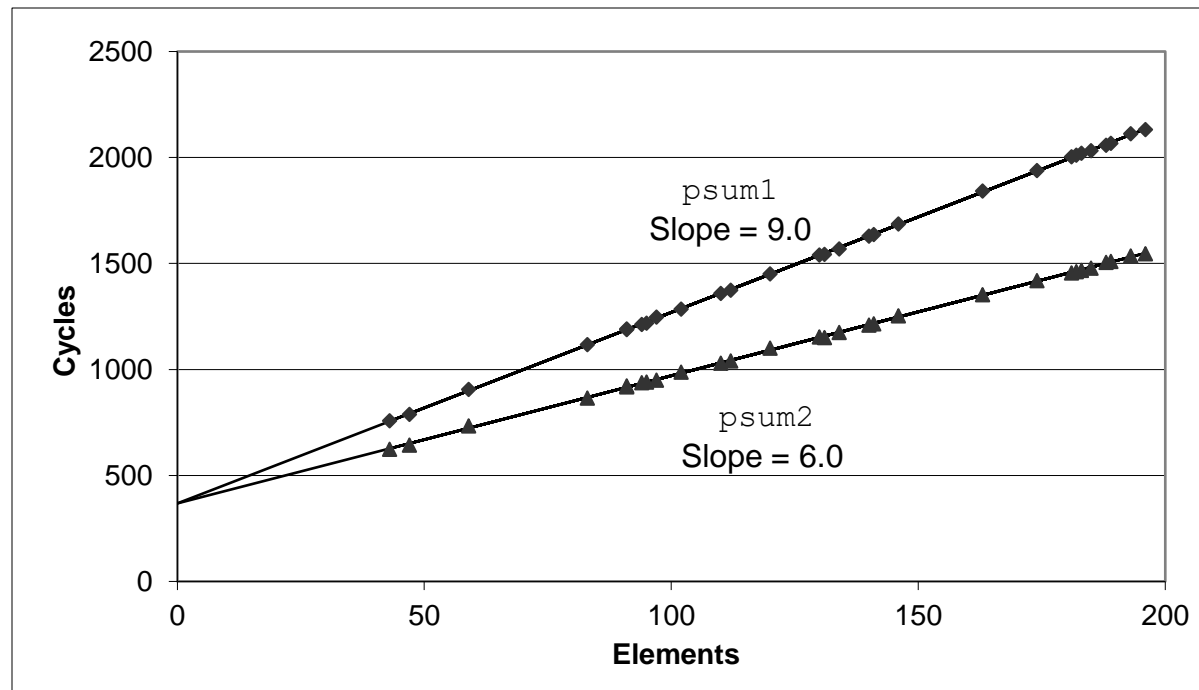
- Use different declarations for `data_t`
- `int`
- `long`
- `float`
- `double`

■ Operations

- Use different definitions of `OP` and `IDENT`
- `+` / `0`
- `*` / `1`

Cycles Per Element (CPE)

- Convenient way to express performance of program that operates on vectors or lists
- Length = n
- In our case: **CPE = cycles per OP**
- **Cycles = CPE * n + Overhead**
 - CPE is slope of line



Benchmark Performance

```

void combine1(vec_ptr 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;
    }
}

```

Compute sum or
product of vector
elements

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine1 unoptimized	22.68	20.02	19.98	20.18
Combine1 -O1	10.12	10.12	10.17	11.14
Combine1 -O3	4.5	4.5	6	7.8

Results in CPE (cycles per element)

Basic Optimizations

```
void combine4(vec_ptr v, data_t *dest)
{
    long i;
    long 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;
}
```

- Move `vec_length` out of loop
- Avoid bounds check on each cycle
- Accumulate in temporary

Effect of Basic Optimizations

```
void combine4(vec_ptr v, data_t *dest)
{
    long i;
    long 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;
}
```

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine1 -O1	10.12	10.12	10.17	11.14
Combine4	1.27	3.01	3.01	5.01

- Eliminates sources of overhead in loop

Loop Unrolling (2x1)

```
void unroll2a_combine(vec_ptr v, data_t *dest)
{
    long length = vec_length(v);
    long limit = length-1;
    data_t *d = get_vec_start(v);
    data_t x = IDENT;
    long i;
    /* Combine 2 elements at a time */
    for (i = 0; i < limit; i+=2) {
        x = (x OP d[i]) OP d[i+1];
    }
    /* Finish any remaining elements */
    for (; i < length; i++) {
        x = x OP d[i];
    }
    *dest = x;
}
```

- Perform 2x more useful work per iteration

Effect of Loop Unrolling

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine4	1.27	3.01	3.01	5.01
Unroll 2x1	1.01	3.01	3.01	5.01
<i>Latency Bound</i>	<i>1.00</i>	<i>3.00</i>	<i>3.00</i>	<i>5.00</i>

■ Helps integer add

- Achieves latency bound

```
x = (x OP d[i]) OP d[i+1];
```

■ Others don't improve. *Why?*

- Sequential dependency

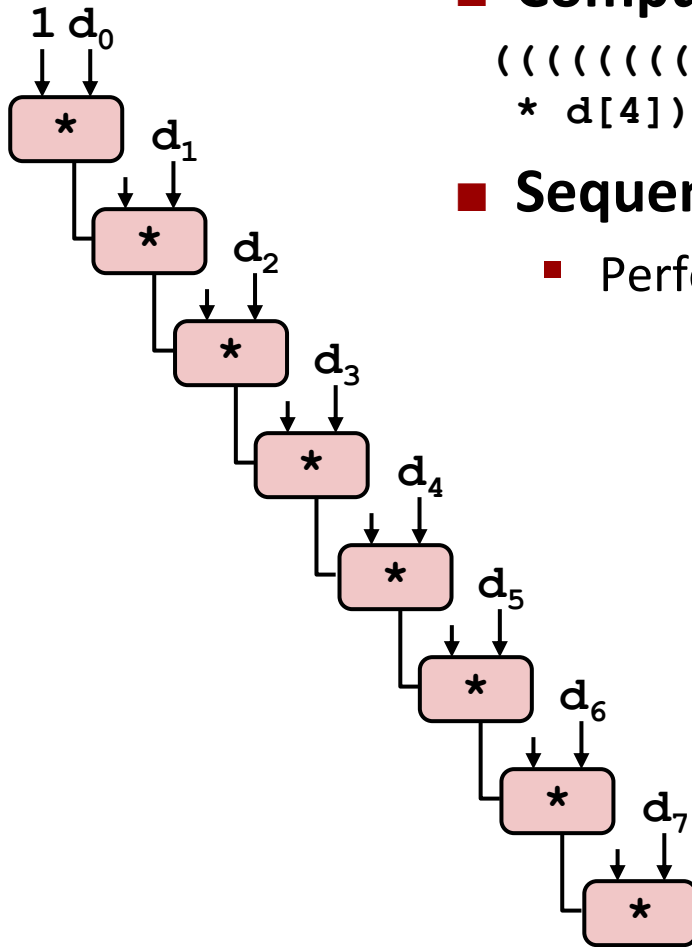
Combine4 = Serial Computation (OP = *)

■ Computation (length=8)

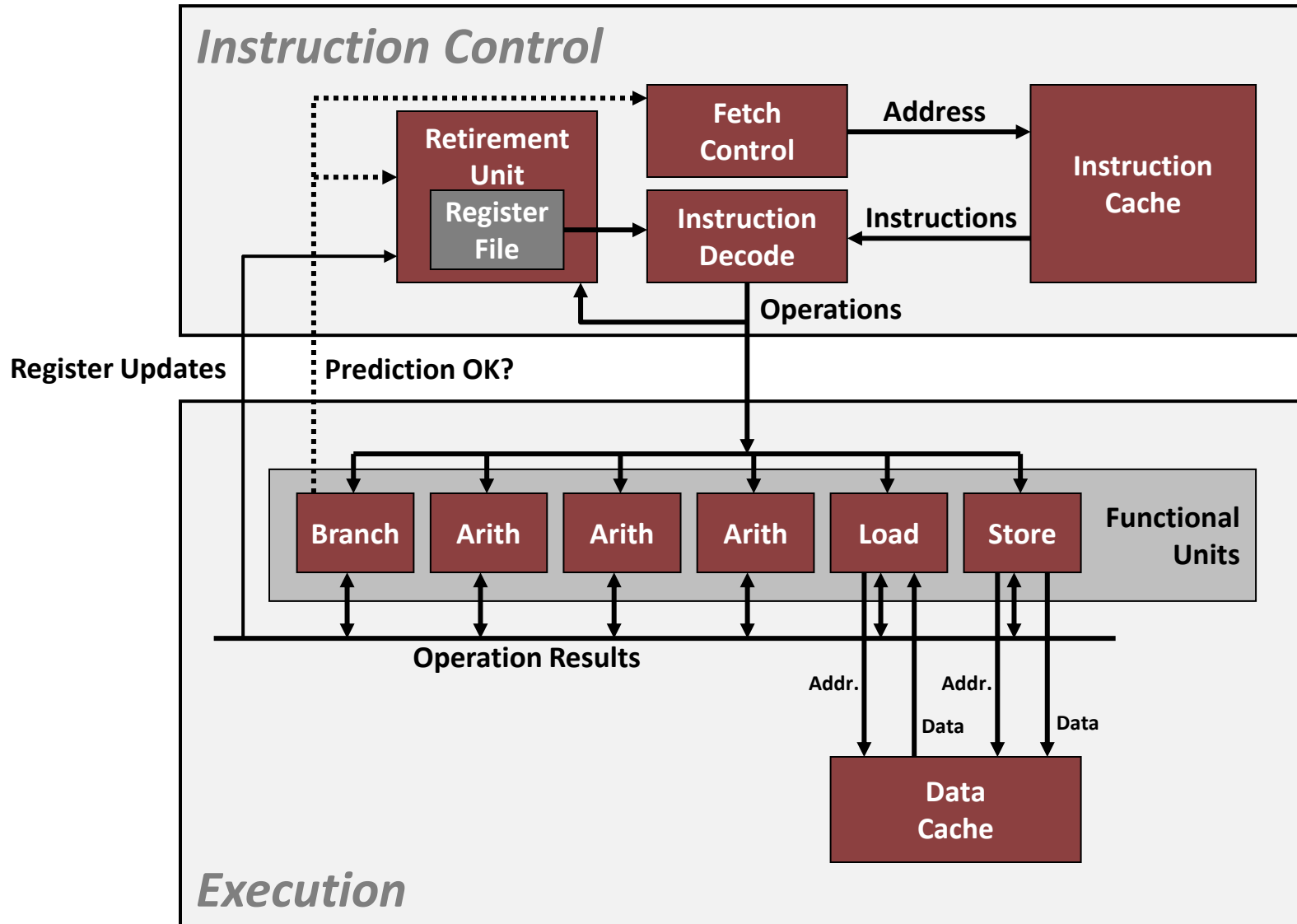
$(((((1 * d[0]) * d[1]) * d[2]) * d[3]) * d[4]) * d[5]) * d[6]) * d[7])$

■ Sequential dependence

- Performance: determined by latency of OP



Modern CPU Design

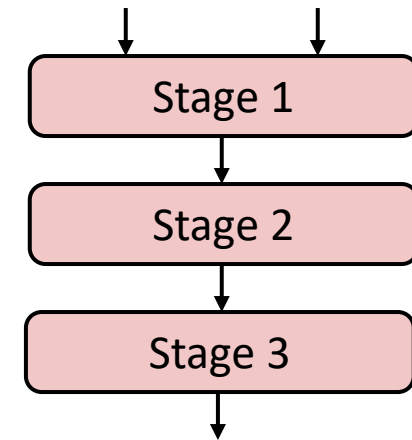


Superscalar Processor

- **Definition:** A superscalar processor can issue and execute *multiple instructions in one cycle*. The instructions are retrieved from a sequential instruction stream and are usually scheduled dynamically.
- **Benefit:** without programming effort, superscalar processor can take advantage of the *instruction level parallelism* that most programs have
- Most modern CPUs are superscalar.
- Intel: since Pentium (1993)

Pipelined Functional Units

```
long mult_eg(long a, long b, long c) {
    long p1 = a*b;
    long p2 = a*c;
    long p3 = p1 * p2;
    return p3;
}
```



Time							
	1	2	3	4	5	6	7
Stage 1	a*b	a*c			p1*p2		
Stage 2		a*b	a*c			p1*p2	
Stage 3			a*b	a*c			p1*p2

- Divide computation into stages
- Pass partial computations from stage to stage
- Stage i can start on new computation once values passed to $i+1$
- E.g., complete 3 multiplications in 7 cycles, even though each requires 3 cycles

Haswell CPU

- 8 Total Functional Units
- **Multiple instructions can execute in parallel**
 - 2 load, with address computation
 - 1 store, with address computation
 - 4 integer
 - 2 FP multiply
 - 1 FP add
 - 1 FP divide
- **Some instructions take > 1 cycle, but can be pipelined**

<i>Instruction</i>	<i>Latency</i>	<i>Cycles/Issue</i>
Load / Store	4	1
Integer Multiply	3	1
Integer/Long Divide	3-30	3-30
Single/Double FP Multiply	5	1
Single/Double FP Add	3	1
Single/Double FP Divide	3-15	3-15

x86-64 Compilation of Combine4

■ Inner Loop (Case: Integer Multiply)

```

.L519:                                # Loop:
    imull  (%rax,%rdx,4), %ecx        # t = t * d[i]
    addq   $1, %rdx                  # i++
    cmpq   %rdx, %rbp                # Compare length:i
    jg     .L519                     # If >, goto Loop

```

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine4	1.27	3.01	3.01	5.01
Latency Bound	1.00	3.00	3.00	5.00

Loop Unrolling with Reassociation (2x1a)

```
void unroll2aa_combine(vec_ptr v, data_t *dest)
{
    long length = vec_length(v);
    long limit = length-1;
    data_t *d = get_vec_start(v);
    data_t x = IDENT;
    long i;
    /* Combine 2 elements at a time */
    for (i = 0; i < limit; i+=2) {
        x = x OP (d[i] OP d[i+1]);
    }
    /* Finish any remaining elements */
    for (; i < length; i++) {
        x = x OP d[i];
    }
    *dest = x;
}
```

Compare to before

$x = (x \text{ OP } d[i]) \text{ OP } d[i+1];$

- Can this change the result of the computation?
- Yes, for FP. *Why?*

Effect of Reassociation

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine4	1.27	3.01	3.01	5.01
Unroll 2x1	1.01	3.01	3.01	5.01
Unroll 2x1a	1.01	1.51	1.51	2.51
<i>Latency Bound</i>	<i>1.00</i>	<i>3.00</i>	<i>3.00</i>	<i>5.00</i>
<i>Throughput Bound</i>	<i>0.50</i>	<i>1.00</i>	<i>1.00</i>	<i>0.50</i>

4 func. units for int +,
2 func. units for load
Why Not .25?

1 func. unit for FP +
3-stage pipelined FP +

2 func. units for FP *,
2 func. units for load
5-stage pipelined FP *

■ Nearly 2x speedup for Int *, FP +, FP *

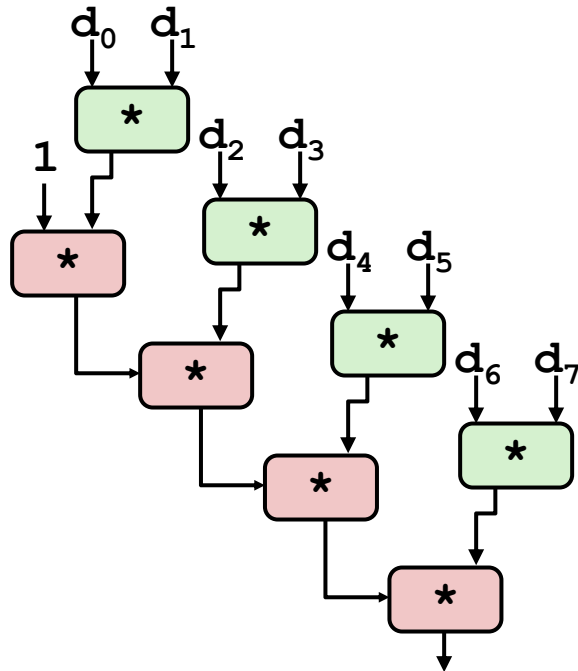
- Reason: Breaks sequential dependency

```
x = x OP (d[i] OP d[i+1]);
```

- Why is that? (next slide)

Reassociated Computation

```
x = x OP (d[i] OP d[i+1]);
```



■ What changed:

- Ops in the next iteration can be started early (no dependency)

■ Overall Performance

- N elements, D cycles latency/op
- $(N/2+1)*D$ cycles:
 $CPE = D/2$

Loop Unrolling with Separate Accumulators (2x2)

```
void unroll2a_combine(vec_ptr v, data_t *dest)
{
    long length = vec_length(v);
    long limit = length-1;
    data_t *d = get_vec_start(v);
    data_t x0 = IDENT;
    data_t x1 = IDENT;
    long i;
    /* Combine 2 elements at a time */
    for (i = 0; i < limit; i+=2) {
        x0 = x0 OP d[i];
        x1 = x1 OP d[i+1];
    }
    /* Finish any remaining elements */
    for (; i < length; i++) {
        x0 = x0 OP d[i];
    }
    *dest = x0 OP x1;
}
```

- Different form of reassociation

Effect of Separate Accumulators

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Combine4	1.27	3.01	3.01	5.01
Unroll 2x1	1.01	3.01	3.01	5.01
Unroll 2x1a	1.01	1.51	1.51	2.51
Unroll 2x2	0.81	1.51	1.51	2.51
<i>Latency Bound</i>	<i>1.00</i>	<i>3.00</i>	<i>3.00</i>	<i>5.00</i>
<i>Throughput Bound</i>	<i>0.50</i>	<i>1.00</i>	<i>1.00</i>	<i>0.50</i>

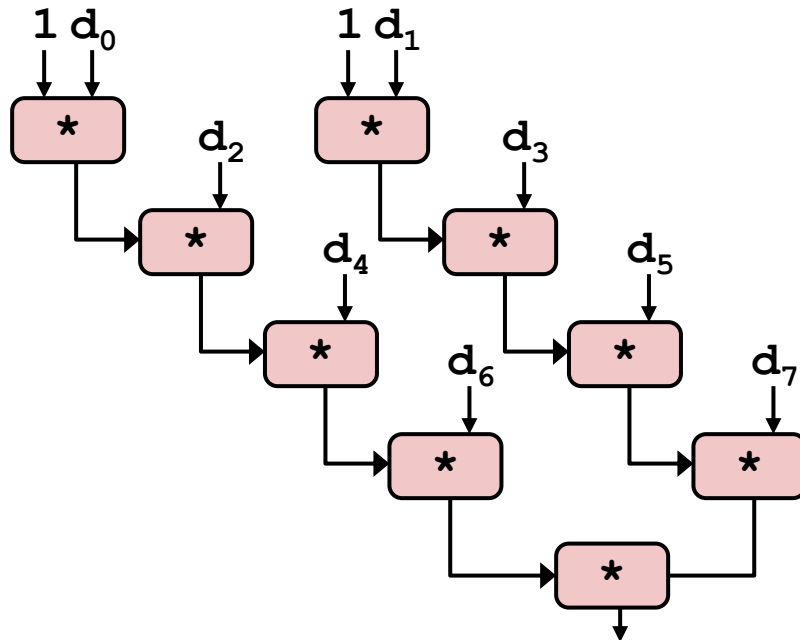
- Int + makes use of two load units

```
x0 = x0 OP d[i];
x1 = x1 OP d[i+1];
```

- 2x speedup (over unroll2) for Int *, FP +, FP *

Separate Accumulators

```
x0 = x0 OP d[i];
x1 = x1 OP d[i+1];
```



■ What changed:

- Two independent “streams” of operations

■ Overall Performance

- N elements, D cycles latency/op
- Should be $(N/2+1)*D$ cycles:
 $CPE = D/2$
- CPE matches prediction!

What Now?

Unrolling & Accumulating

■ Idea

- Can unroll to any degree L
- Can accumulate K results in parallel
- L must be multiple of K

■ Limitations

- Diminishing returns
 - Cannot go beyond throughput limitations of execution units
- Large overhead for short lengths
 - Finish off iterations sequentially

Unrolling & Accumulating: Double *

■ Case

- Intel Haswell
- Double FP Multiplication
- Latency bound: 5.00. Throughput bound: 0.50

<i>Accumulators</i>	FP *	Unrolling Factor L							
	K	1	2	3	4	6	8	10	12
	1	5.01	5.01	5.01	5.01	5.01	5.01	5.01	
	2		2.51		2.51		2.51		
	3			1.67					
	4				1.25		1.26		
	6					0.84			0.88
	8						0.63		
	10							0.51	
	12								0.52

Achievable Performance

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Best	0.54	1.01	1.01	0.52
<i>Latency Bound</i>	<i>1.00</i>	<i>3.00</i>	<i>3.00</i>	<i>5.00</i>
<i>Throughput Bound</i>	<i>0.50</i>	<i>1.00</i>	<i>1.00</i>	<i>0.50</i>

- Limited only by throughput of functional units
- Up to 42X improvement over original, unoptimized code

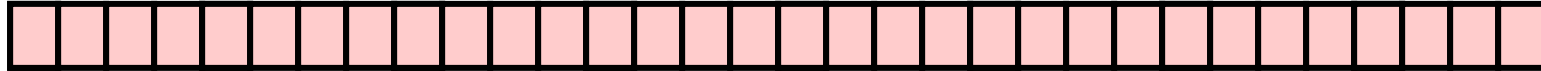
Can we do even better?

Programming with AVX2

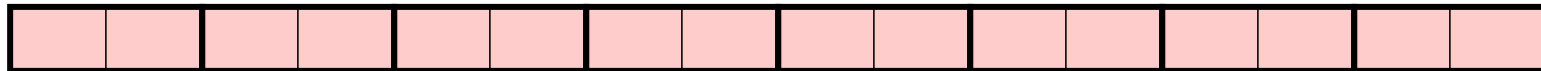
YMM Registers

■ 16 total, each 32 bytes

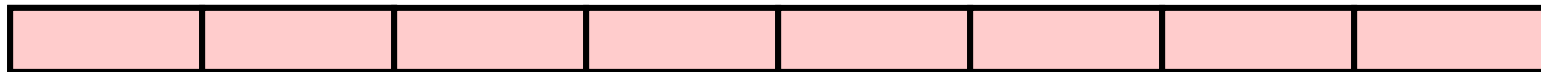
■ 32 single-byte integers



■ 16 16-bit integers



■ 8 32-bit integers



■ 8 single-precision floats



■ 4 double-precision floats



■ 1 single-precision float



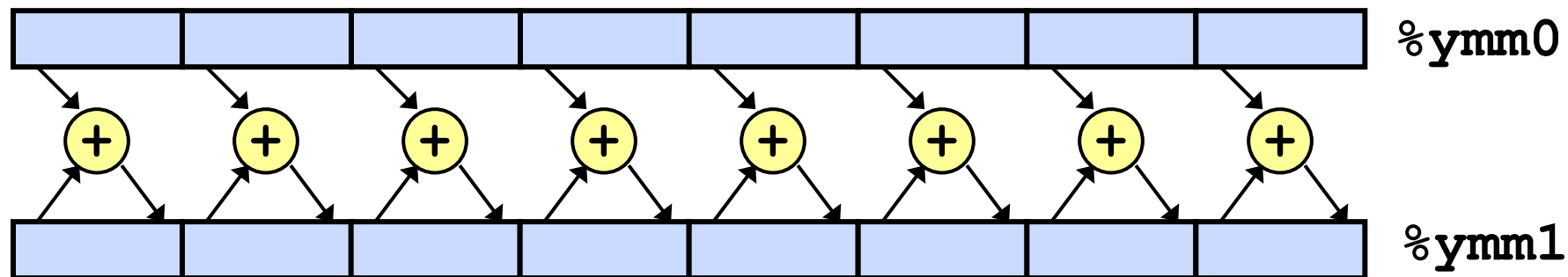
■ 1 double-precision float



SIMD Operations

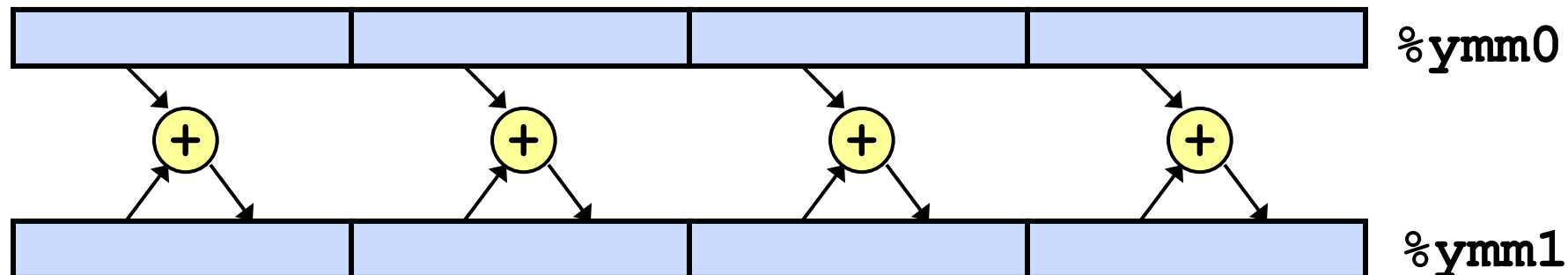
■ SIMD Operations: Single Precision

vaddps %ymm0, %ymm1, %ymm1



■ SIMD Operations: Double Precision

vaddpd %ymm0, %ymm1, %ymm1



Using Vector Instructions

Method	Integer		Double FP	
Operation	Add	Mult	Add	Mult
Scalar Best	0.54	1.01	1.01	0.52
Vector Best	0.06	0.24	0.25	0.16
<i>Latency Bound</i>	<i>0.50</i>	<i>3.00</i>	<i>3.00</i>	<i>5.00</i>
<i>Throughput Bound</i>	<i>0.50</i>	<i>1.00</i>	<i>1.00</i>	<i>0.50</i>
<i>Vec Throughput Bound</i>	<i>0.06</i>	<i>0.12</i>	<i>0.25</i>	<i>0.12</i>

■ Make use of AVX Instructions

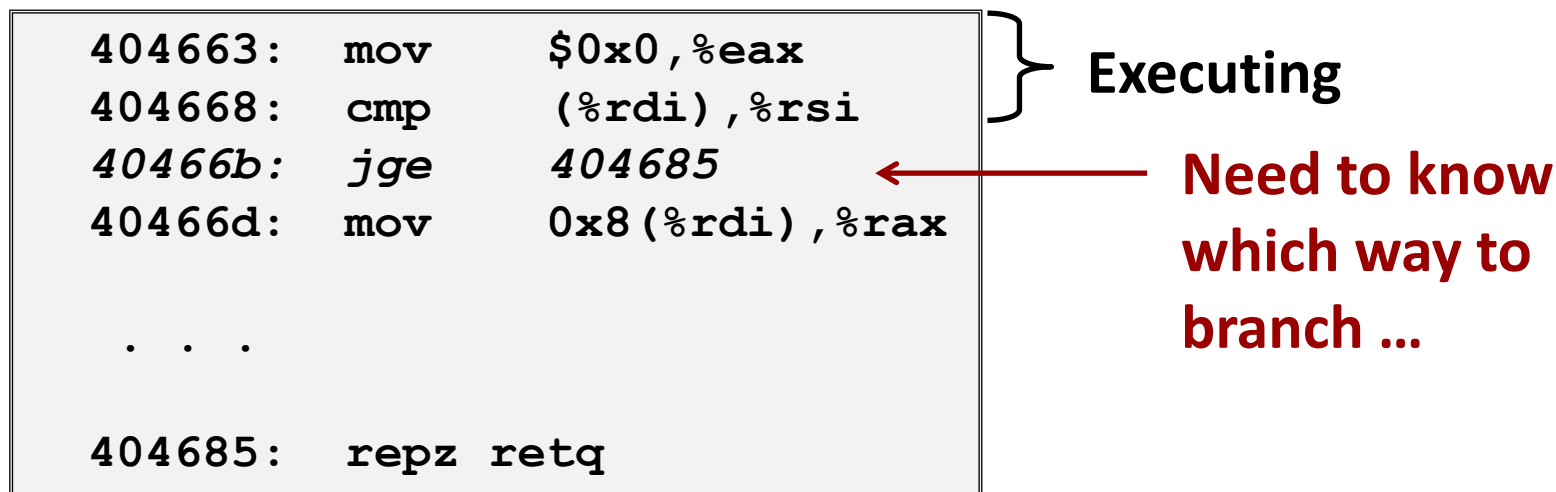
- Parallel operations on multiple data elements
- See Web Aside OPT:SIMD on CS:APP web page

Today

- Machine-Dependent Optimizations
- Instruction-Level Parallelism
- **Branch Predictions**

Branches Are A Challenge

- **Instruction Control Unit** must work well ahead of **Execution Unit** to generate enough operations to keep EU busy



If the CPU has to wait for the result of the `cmp` before continuing to fetch instructions, may waste tens of cycles doing nothing!

Branch Prediction

■ *Guess which way branch will go*

- Begin executing instructions at predicted position
- But don't actually modify register or memory data

```
404663:  mov    $0x0,%eax
404668:  cmp    (%rdi),%rsi
40466b:  jge    404685
40466d:  mov    0x8(%rdi),%rax
. . .
404685:  repz   retq
```

Predict Taken

**Continue
Fetching
Here**

Branch Prediction Through Loop

```
401029:  mulsd    (%rdx), %xmm0, %xmm0
40102d:  add      $0x8, %rdx
401031:  cmp      %rax, %rdx
401034:  jne      401029
```

i = 98

Assume
array length = **100**

Predict Taken (OK)

```
401029:  mulsd    (%rdx), %xmm0, %xmm0
40102d:  add      $0x8, %rdx
401031:  cmp      %rax, %rdx
401034:  jne      401029
```

i = 99

Predict Taken
(Oops)

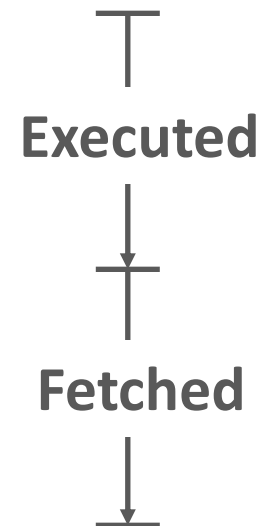
```
401029:  mulsd    (%rdx), %xmm0, %xmm0
40102d:  add      $0x8, %rdx
401031:  cmp      %rax, %rdx
401034:  jne      401029
```

i = 100

Read
invalid
location

```
401029:  mulsd    (%rdx), %xmm0, %xmm0
40102d:  add      $0x8, %rdx
401031:  cmp      %rax, %rdx
401034:  jne      401029
```

i = 101



Branch Misprediction Invalidation

```
401029:  mulsd    (%rdx), %xmm0, %xmm0
40102d:  add      $0x8, %rdx
401031:  cmp      %rax, %rdx
401034:  jne      401029
```

i = 98

Assume
array length = **100**

Predict Taken (OK)

```
401029:  mulsd    (%rdx), %xmm0, %xmm0
40102d:  add      $0x8, %rdx
401031:  cmp      %rax, %rdx
401034:  jne      401029
```

i = 99

Predict Taken
(Oops)

```
401029:  mulsd    (%rdx), %xmm0, %xmm0
40102d:  add      $0x8, %rdx
401031:  cmp      %rax, %rdx
401034:  jne      401029
```

i = 100

Invalidate

```
401029:  mulsd    (%rdx), %xmm0, %xmm0
40102d:  add      $0x8, %rdx
401031:  cmp      %rax, %rdx
401034:  jne      401029
```

i = 101

Branch Misprediction Recovery

```
401029:  mulsd    (%rdx), %xmm0, %xmm0
```

```
40102d:  add      $0x8, %rdx
```

```
401031:  cmp      %rax, %rdx
```

```
401034:  jne      401029
```

```
401036:  jmp      401040
```

```
. . .
```

```
401040:  movsd    %xmm0, (%r12)
```

i = 99

Definitely not taken

Reload
Pipeline

■ Performance Cost

- Multiple clock cycles on modern processor
- Can be a major performance limiter

Branch Prediction Numbers

■ A simple heuristic:

- Backwards branches are often loops, so predict taken
- Forwards branches are often ifs, so predict not taken
- >95% prediction accuracy just with this!

■ Fancier algorithms track behavior of each branch

- Subject of ongoing research
- 2011 record (<https://www.jilp.org/jwac-2/program/JWAC-2-program.htm>): 34.1 mispredictions per 1000 instructions
- Current research focuses on the remaining handful of “impossible to predict” branches (strongly data-dependent, no correlation with history)
 - e.g. https://hps.ece.utexas.edu/pub/PruettPatt_BranchRunahead.pdf

Optimizing for Branch Prediction

■ Loop Unrolling

```
void unroll2a_combine(vec_ptr v, data_t *dest)
{
    long length = vec_length(v);
    long limit = length-1;
    data_t *d = get_vec_start(v);
    data_t x = IDENT;
    long i;
    /* Combine 2 elements at a time */
    for (i = 0; i < limit; i+=2) {
        x = (x OP d[i]) OP d[i+1];
    }
    /* Finish any remaining elements */
    for (; i < length; i++) {
        x = x OP d[i];
    }
    *dest = x;
}
```


Optimizing for Branch Prediction

■ Transform Branches

```
for (int c=0; c < size; ++c)
{
    if (data[c] >= 128)
        sum += data[c];
}
```

```
.L4:
    mov    rdx, QWORD PTR [rax]
    cmp    rdx, 127
    jbe    .L3
    add    rcx, rdx
    mov    edi, 1
.L3:
    next loop
```

```
int t = (data[c] - 128) >> 31;
sum += ~t & data[c];
```

```
.L3:
    mov    rcx, QWORD PTR [rdx]
    lea    rax, [rcx-128]
    shr    rax, 31
    not    eax
    cdq    eax
    and    rax, rcx
    add    rsi, rax
    add    rdx, 8
    cmp    rdx, rdi
    jne    .L3
```

Optimizing for Branch Prediction

■ Conditional Moves

```
int absdiff(int x, int y) {
    int result;
    if (x > y) result = x - y;
    else result = y - x;
    return result;
}
```

```
absdiff:
    mov     eax, DWORD PTR [rbp-20]
    cmp     eax, DWORD PTR [rbp-24]
    jle     .L2
    mov     eax, DWORD PTR [rbp-20]
    sub     eax, DWORD PTR [rbp-24]
    mov     DWORD PTR [rbp-4], eax
    jmp     .L3
.L2:
    mov     eax, DWORD PTR [rbp-24]
    sub     eax, DWORD PTR [rbp-20]
    mov     DWORD PTR [rbp-4], eax
.L3:
    mov     eax, DWORD PTR [rbp-4]
```

```
absdiff:
    mov     edx, edi
    sub     edx, esi
    mov     eax, esi
    sub     eax, edi
    cmp     edi, esi
    cmovg   eax, edx
```

Not Always a
Good Idea

Optimizing for Branch Prediction

■ Make Branch More Predictable

```
for (int c=0; c < size; ++c) {
    if (data[c] >= 128)
        sum += data[c];
}
```

T = branch taken
N = branch not taken

```
data[] = 226, 185, 125, 158, 198, 144, 217, 79, 202, 118, 14, ...
branch =  T,   T,   N,   T,   T,   T,   T,   N,   T,   N,   N, ...

      = TTNTTTTNTNNTTT ...    (completely random)
```

```
data[] = 0, 1, 2, 3, 4, ... 126, 127, 128, 129, 130, ... 250, ...
branch = N  N  N  N  N  ...   N   N   T   T   T   ...   T, ...

      = NNNNNNNNNNNN ... NNNNNNNNTTTTTTTTTT ... (easy to predict)
```

Going Further

- **Compiler optimizations are an easy gain**
 - 20 CPE down to 3-5 CPE
- **With careful hand tuning and computer architecture knowledge**
 - 4-16 elements per cycle
 - Newest compilers are closing this gap

Summary: Getting High Performance

- **Good compiler and flags**
- **Don't do anything sub-optimal**
 - Watch out for hidden algorithmic inefficiencies
 - Write compiler-friendly code
 - Watch out for optimization blockers:
procedure calls & memory references
 - Look carefully at innermost loops (where most work is done)
- **Tune code for machine**
 - Exploit instruction-level parallelism
 - Avoid unpredictable branches
 - Make code cache friendly