

Advanced Systems Lab

Spring 2024

Lecture: Optimization for Instruction-Level Parallelism

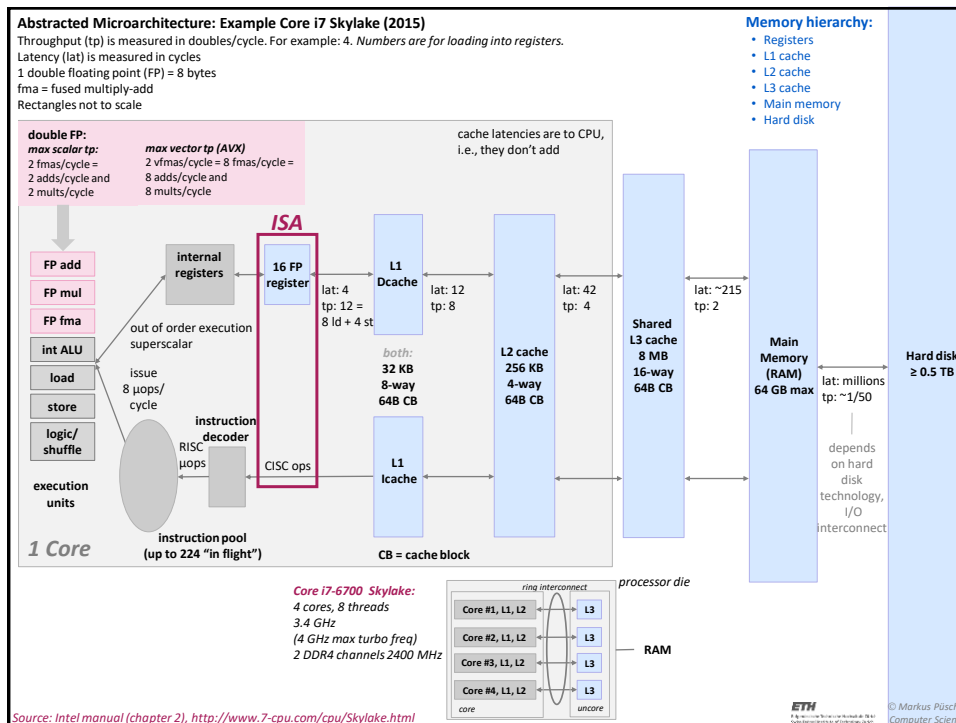
Instructor: Markus Püschel

TA: Tommaso Pegolotti, several more



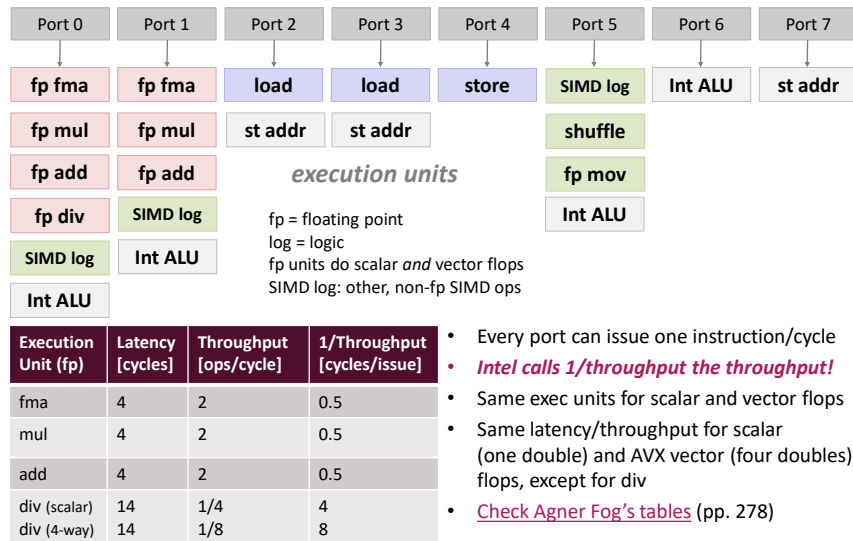
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Execution Units and Ports (Skylake)



Source: Intel manual (Table C-8. 256-bit AVX Instructions, Table 2-1. Dispatch Port and Execution Stacks of the Skylake Microarchitecture, Figure 2-1. CPU Core Pipeline Functionality of the Skylake Microarchitecture)

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How To Make Code Faster?

It depends! First high-level approaches:

Memory bound: Reduce memory traffic

- Reduce cache misses
- Compress data

Compute bound: Keep floating point units busy

- Reduce cache misses, register spills
- Instruction level parallelism (ILP)
- Vectorization

Next: Optimizing for ILP (an example)

Chapter 5 in *Computer Systems: A Programmer's Perspective, 2nd edition*, Randal E. Bryant and David R. O'Hallaron, Addison Wesley 2010

Part of these slides are adapted from the course associated with this book

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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: Superscalar processors can take advantage of *instruction level parallelism (ILP)* that many programs have.

Most desktop/laptop CPUs since about 1998 are superscalar

Intel: since Pentium Pro

Simple embedded processors are usually not superscalar

Deep pipelines also require ILP (explained today).

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ILP

Code

```
t2 = t0 + t1  
t5 = t4 * t3  
t6 = t2 + t5
```

can be executed in parallel
and in any order

t2 = t0 + t1

t5 = t4 * t3

Dependencies

t6 = t2 + t5

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Hard Bounds: Coffee Lake and Haswell

Coffee Lake (2017, Skylake family)

	latency	1/throughput
FP Add	4	0.5
FP Mul	4	0.5
Int Add	1	0.5
Int Mul	3	1

} **blackboard**

Haswell (2013)

	latency	1/throughput
FP Add	3	1
FP Mul	5	0.5
Int Add	1	0.5
Int Mul	3	1

More precisely (Int Add):

- tp = 2 if one operand is from mem
- tp = 4 if all operands are in register

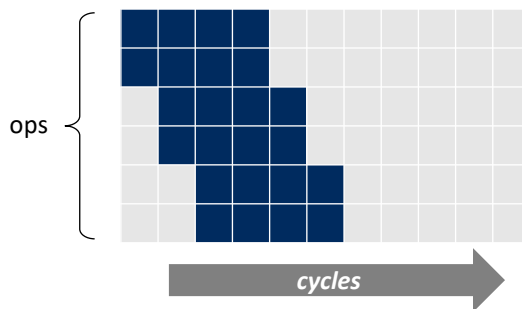
Only the ALU at Port 1 does Int Mults

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Coffee Lake

	latency	1/throughput
FP Mul	4	0.5



Throughput: 2/cycle

How many cycles at least for n mults?

- $\text{ceil}(n/2)$ (considering only throughput)
- $\text{ceil}(n/2) + 3$ (considering latency and throughput)

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Example Computation: Reduction

```
void reduce(vec_ptr 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;
}
```

$d[0] \text{ OP } d[1] \text{ OP } d[2] \text{ OP } \dots \text{ OP } d[\text{length}-1]$

data_t: double or int

OP: + or *

IDENT: 0 or 1

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Runtime of Reduce (Coffee Lake)

```
void reduce(vec_ptr 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;
}
```

Measured cycles per OP

Method	Int (add/mult)	Float (add/mult)		
reduce	1.29	2.95	3.91	3.91
bound	0.5	1.0	0.5	0.5

Questions:

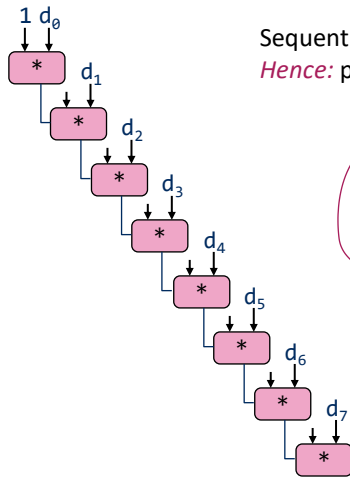
- Explain red row
- Explain gray row

This and all following measurements: gcc -O3 -mavx2 -fno-tree-vectorize

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Reduce = Serial Computation (here: *)



Sequential dependence = no ILP!

Hence: performance determined by latency of OP!

Method	Int (add/mult)	Float (add/mult)
reduce	1.29	2.95
bound	0.5	1.0

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Loop Unrolling

```
void unroll2(vec_ptr v, data_t *dest)
{
    int length = vec_length(v);
    int limit = length-1;
    data_t *d = get_vec_start(v);
    data_t x = IDENT;
    int 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

How does the runtime change?

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Effect of Loop Unrolling

Method	Int (add/mult)		Float (add/mult)	
combine4	1.29	2.95	3.91	3.91
unroll2	1.0	2.93	3.90	3.91
bound	0.5	1.0	0.5	0.5



Helps integer sum a bit

Others don't improve. *Why?*

- Still sequential dependency

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

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Loop Unrolling with Separate Accumulators

```
void unroll2_sa(vec_ptr v, data_t *dest)
{
    int length = vec_length(v);
    int limit = length-1;
    data_t *d = get_vec_start(v);
    data_t x0 = IDENT;
    data_t x1 = IDENT;
    int 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;
}
```

Effect on runtime?

Can this change the result of the computation?

Floating point: yes!

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Effect of Separate Accumulators

Method	Int (add/mult)		Float (add/mult)	
combine4	1.29	2.95	3.91	3.91
unroll2	1.0	2.93	3.90	3.91
unroll2-sa	0.8	1.49	1.96	1.97
bound	0.5	1.0	0.5	0.5

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

- Breaks sequential dependency

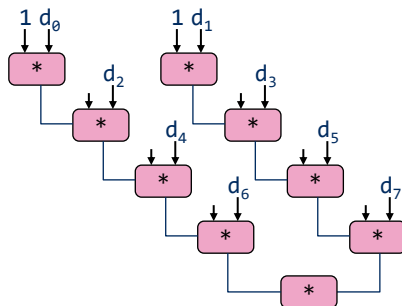
```
x0 = x0 OP d[i];
x1 = x1 OP d[i+1];
```

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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:
cycles per OP $\approx D/2$

What Now?

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Unrolling & Accumulating

Idea

- Use K accumulators
- Increase K until best performance reached
- Need to unroll by L , K divides L

Limitations

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

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Unrolling & Accumulating: FP *

Coffee Lake: FP multiplication

- $1/\text{Throughput} = \text{cycles/issue} = 0.5$
- $\text{Throughput} = 2$
- $\text{Latency} = 4$

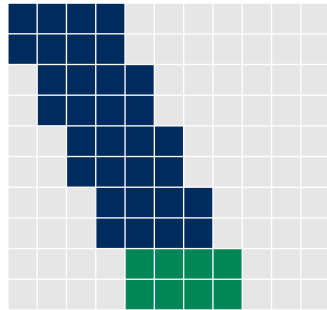
FP64 *	Unrolling Factor L							
K	1	2	3	4	6	8	10	12
1	3.91	3.91		3.91		3.91		
2		1.97		1.97		1.96		
3			1.32		1.32			
4				1.00		1.0		
6					0.70			0.70
8						0.56		
10							0.54	
12								0.54

Why 8?

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Why 8?



Those have to be independent

Based on this insight: $K = \text{\#accumulators}$
 $= \text{ceil}(\text{latency} / \text{cycles per issue})$
 $= \text{ceil}(\text{latency} * \text{throughput})$

Here (FP Mult): $K = \text{ceil}(4 / 0.5) = \text{ceil}(4 * 2) = 8$

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Unrolling & Accumulating: FP +

Coffee Lake: FP addition

- *Throughput* = 2
- *Latency* = 4

FP64 +	Unrolling Factor L								
Accumulators	K	1	2	3	4	6	8	10	12
1		3.91	3.90		3.90		3.90		
2			1.96		1.96		1.96		
3				1.32		1.32			
4					1.00		1.00		
6						0.70			0.70
8							0.56		
10								0.54	
12									0.54

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Unrolling & Accumulating: Int *

Coffee Lake: Int multiplication

- *Throughput = 1*
- *Latency = 3*

Accumulators	Int *	Unrolling Factor L							
	K	1	2	3	4	6	8	10	12
1		2.94	2.94		2.93		2.93		
2			1.49		1.49		1.49		
3				1.32		1.32			
4					1.01		1.01		
6						1.01			1.00
8							1.01		
10								1.01	
12									1.01

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Unrolling & Accumulating: Int +

Coffee Lake: Int multiplication

- *Throughput = 2*
- *Latency = 1*

Accumulators	Int +	Unrolling Factor L							
	K	1	2	3	4	6	8	10	12
1		1.29	1.00		1.00		1.00		
2			0.80		0.58		0.52		
3				0.69		0.52			
4					0.57		0.52		
6						0.52			0.52
8							0.52		
10								0.52	
12									0.52

Interesting question: what exactly happens here?

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Coffee Lake vs. Haswell: FP +

FP64 +	Unrolling Factor L							
K	1	2	3	4	6	8	10	12
1	3.91	3.90		3.90		3.90		
2		1.96		1.96		1.96		
3			1.32		1.32			
4				1.00		1.00		
6					0.70			0.70
8						0.56		
10							0.54	
12								0.54

Coffee Lake:
 Latency = 4
 Throughput = 2

FP64 +	Unrolling Factor L							
K	1	2	3	4	6	8	10	12
1	2.95	2.95		2.95		2.95		
2		1.49		1.49		1.49		
3			1.00		1.00			
4				1.01		1.01		
6					1.01			1.01
8						1.00		
10							1.01	
12								1.01

Haswell:
 Latency = 3
 Throughput = 1

Says something about porting processor-tuned code

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Apple M1: FP +

FP64 +	Unrolling Factor L							
K	1	2	3	4	6	8	10	12
1	2.87	2.83		2.75		2.75		
2		1.38		1.39		1.26		
3			0.93		0.93			
4				0.71		0.71		
6					0.49			0.49
8						0.38		
10							0.32	
12								0.31

Firestorm:
 Latency = 3
 Throughput = 4

about 9x speedup!

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Summary (ILP)

Deep pipelines and multiple ports require ILP for good performance

ILP may have to be made explicit in program

Potential blockers for compilers

- *Reassociation changes result (floating point)*
- *Too many choices, no good way of deciding*

Unrolling

- *By itself does usually nothing (branch prediction works usually well)*
- *But may be needed to enable additional transformations (here: reassociation)*

How to program this example?

- *Solution 1: program generator generates alternatives and picks best*
- *Solution 2: use model based on latency and throughput*

We have seen how to obtain runtime bounds based on throughput and runtime estimates considering latency. Can be done on arbitrary computation DAGs.

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