

# Advanced Systems Lab

Spring 2022

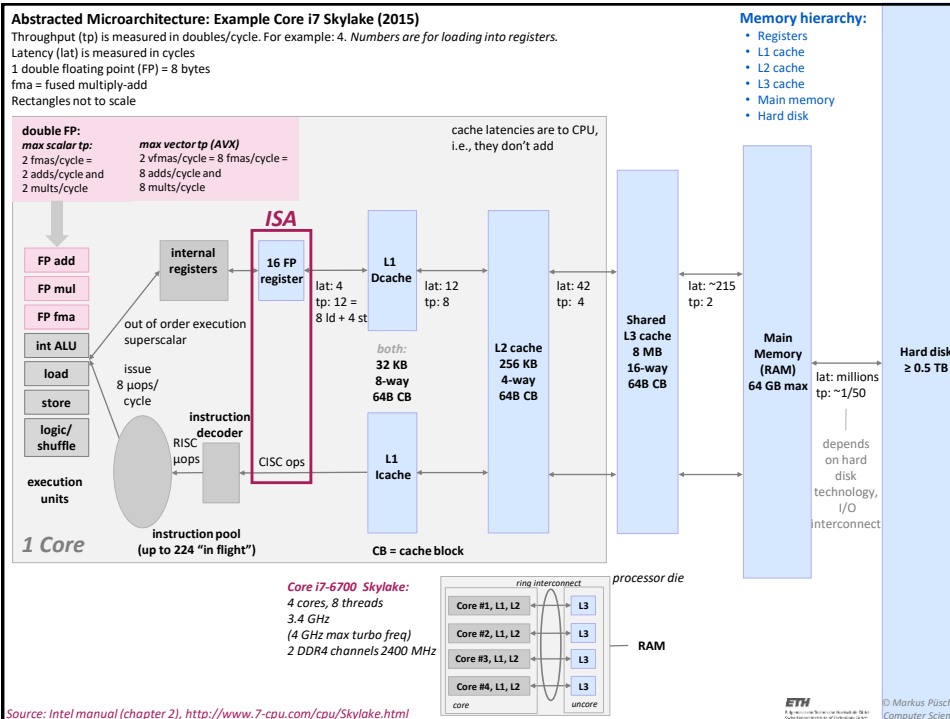
Lecture: Optimization for Instruction-Level Parallelism

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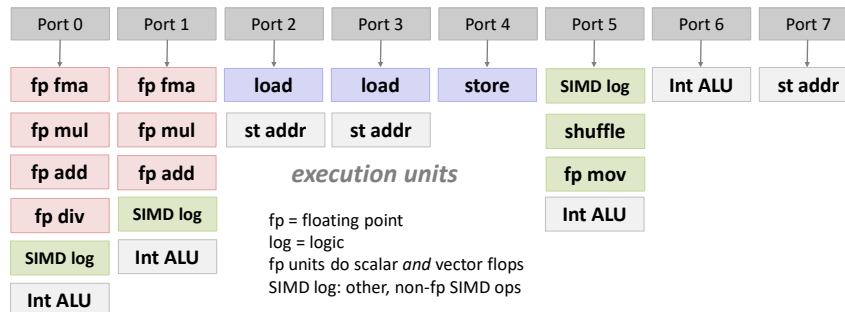
TA: Joao Rivera, several more



Eidgenössische Technische Hochschule Zürich  
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## Mapping of execution units to ports



Execution Unit (fp)	Latency [cycles]	Throughput [ops/cycle]	Gap [cycles/issue]
fma	4	2	0.5
mul	4	2	0.5
add	4	2	0.5
div (scalar)	14	1/4	4
div (4-way)	14	1/8	8

- Every port can issue one instruction/cycle
- Gap = 1/throughput
- **Intel calls gap the throughput!**
- Same exec units for scalar and vector flops
- Same latency/throughput for scalar (one double) and AVX vector (four doubles) flops, except for div

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)

## How To Make Code Faster?

It depends!

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, 2<sup>nd</sup> 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.

Deep pipelines also require ILP (explained today).

Most CPUs since about 1998 are superscalar

Intel: since Pentium Pro

Simple embedded processors are usually not superscalar

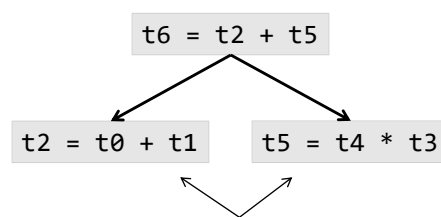
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# ILP

## Code

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

## Dependencies



can be executed in parallel  
and in any order

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# Hard Bounds: Coffee Lake and Haswell

## Coffee Lake

	latency	1/tp = gap
FP Add	4	0.5
FP Mul	4	0.5
Int Add	1	0.5
Int Mul	3	1

} **blackboard**

## Haswell

	latency	1/tp = gap
FP Add	3	1
FP Mul	5	0.5
Int Add	1	0.5
Int Mul	3	1

More precisely (Int Add):

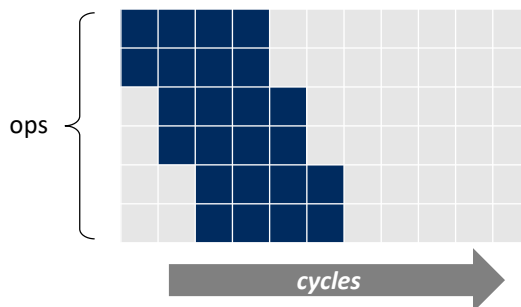
- tp = 2 if one operand is from mem
- tp = 4 if all operands in register

Only the ALU at Port 1 does Int Mults

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## Coffee Lake

	latency	1/tp = gap
FP Mul	4	0.5



Throughput  $tp = 2/\text{cycle}$

Gap =  $1/tp = 1/2$  cycles/issue

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

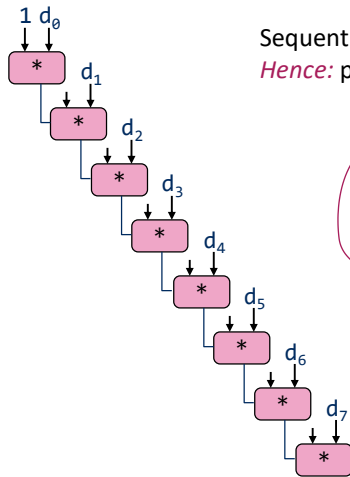
Questions:

- 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 \*

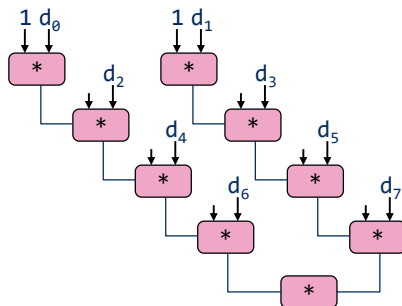
- Breaks sequential dependency

```
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

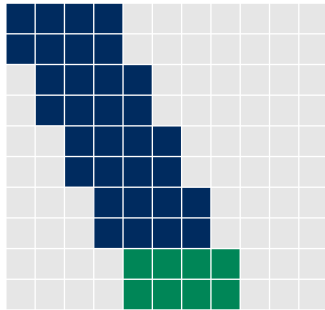
- Gap = cycles/issue = 0.5
- Latency = 4

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

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## Unrolling & Accumulating: FP +

Coffee Lake: FP addition

- $Gap = \text{cycles/issue} = 0.5$
- $Latency = 4$

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

- $Gap = cycles/issue = 1$
- $Latency = 3$

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

- $Gap = cycles/issue = 0.5$
- $Latency = 1$

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  
Gap = 0.5

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  
Gap = 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  
Gap = 0.25

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## Summary (ILP)

Deep pipelines and multiple ports require ILP for good throughput

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*

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