

# “Simple” performance modeling: The Roofline Model

## Loop-based performance modeling: Execution vs. data transfer

R.W. Hockney and I.J. Curington:  $f_{1/2}$ : A parameter to characterize memory and communication bottlenecks. *Parallel Computing* 10, 277-286 (1989). DOI: [10.1016/0167-8191\(89\)90100-2](https://doi.org/10.1016/0167-8191(89)90100-2)

W. Schönauer: Scientific Supercomputing: Architecture and Use of Shared and Distributed Memory Parallel Computers. Self-edition (2000)

S. Williams: Auto-tuning Performance on Multicore Computers. UCB Technical Report No. UCB/EECS-2008-164. PhD thesis (2008)



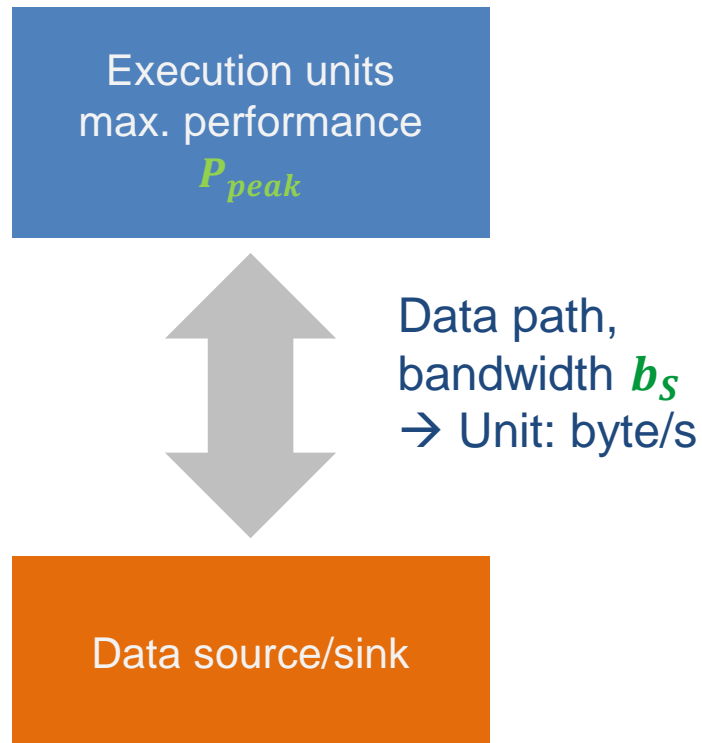
# Analytic white-box performance models

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An analytic white-box performance model is a **simplified mathematical description** of the **hardware** and its interaction with **software**. It is able to **predict the runtime/performance** of code from “**first principles**.”

# A simple performance model for loops

Simplistic view of the hardware:



Simplistic view of the software:

```
! may be multiple levels
do i = 1,<sufficient>
  <complicated stuff doing
    N flops causing
    V bytes of data transfer>
enddo
```

Computational intensity  $I = \frac{N}{V}$   
→ Unit: flop/byte

# Naïve Roofline Model

How fast can tasks be processed?  $P$  [flop/s]

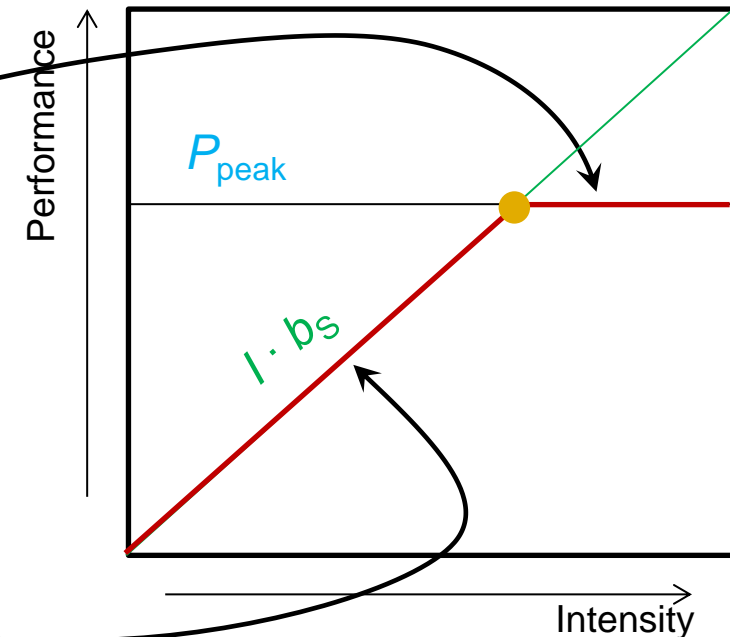
The bottleneck is either

- The execution of work:  $P_{\text{peak}}$  [flop/s]
- The data path:  $I \cdot b_S$  [flop/byte x byte/s]

$$P = \min(P_{\text{peak}}, I \cdot b_S)$$

This is the “Naïve Roofline Model”

- High intensity:  $P$  limited by execution
- Low intensity:  $P$  limited by data transfer
- “Knee” at  $P_{\text{peak}} = I \cdot b_S$ :  
Best use of resources
- Roofline is an “optimistic” model  
(think “light speed”)



# The Roofline Model in computing – Basics

Apply the naive Roofline model in practice

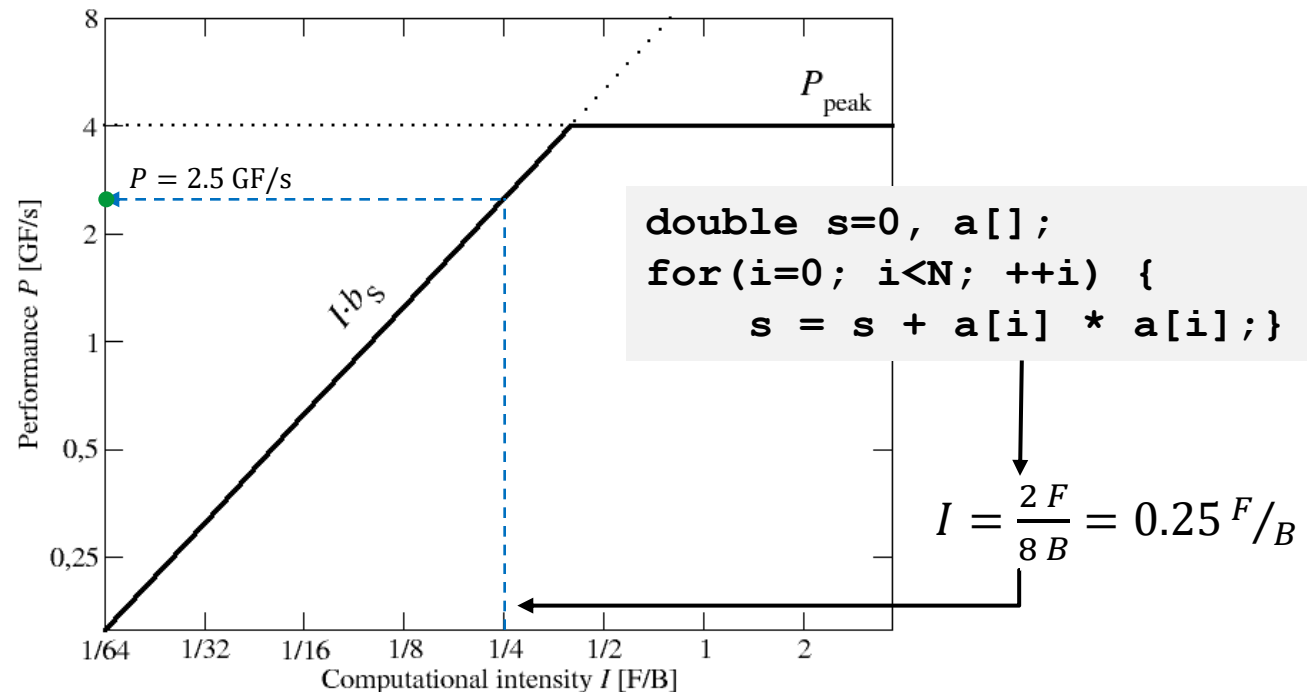
- Machine parameter #1: Peak performance:  $P_{peak} \left[ \frac{F}{s} \right]$
  - Machine parameter #2: Memory bandwidth:  $b_S \left[ \frac{B}{s} \right]$
  - Code characteristic: Computational intensity:  $I \left[ \frac{F}{B} \right]$
- Machine model
- Application model

Machine properties:

$$P_{peak} = 4 \frac{GF}{s}$$

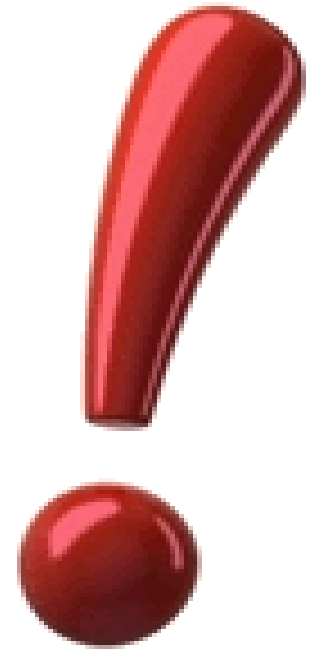
$$b_S = 10 \frac{GB}{s}$$

Application property:  $I$



# Prerequisites for the Roofline Model

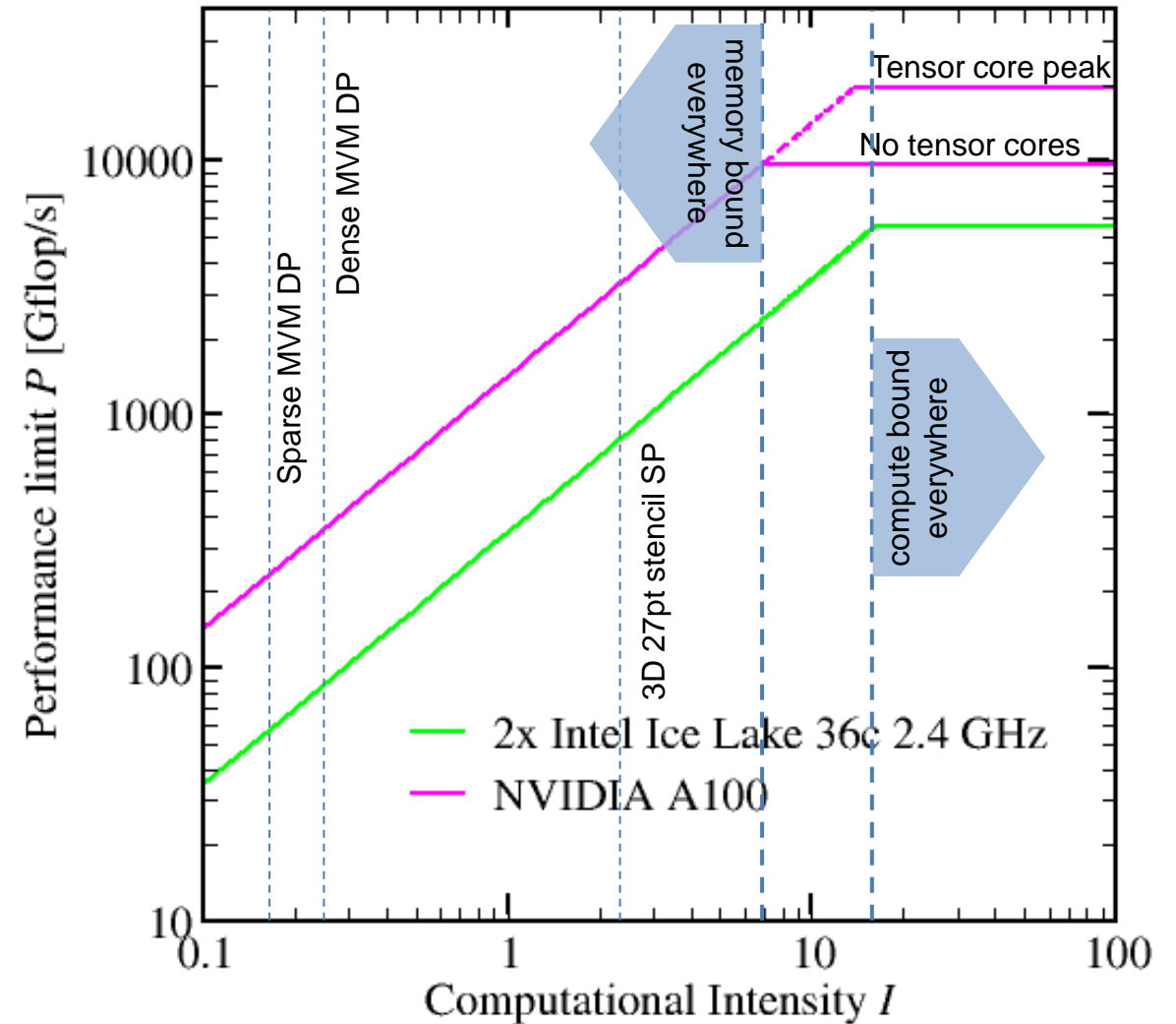
- Data transfer and core execution overlap perfectly!
  - Either the limit is core execution or it is data transfer
- Slowest limiting factor “wins”; all others are assumed to have no impact
  - If two bottlenecks are “close,” no interaction is assumed
- Data access latency is ignored, i.e. perfect streaming mode
  - Achievable bandwidth is the limit
- Chip must be able to saturate the bandwidth bottleneck(s)
  - Always model the full chip



# Roofline for architecture and code comparison

With Roofline, we can

- Compare capabilities of different machines
- Compare performance expectations for different loops
- Roofline always provides upper bound – but is it realistic?
  - Simple case: Loop kernel has loop-carried dependencies → cannot achieve peak
  - Other bandwidth bottlenecks may apply



# A refined Roofline Model

1.  $P_{\max}$  = Applicable peak performance of a loop, assuming that data comes from the level 1 cache (this is **not necessarily**  $P_{\text{peak}}$ )  
→ e.g.,  $P_{\max} = 176$  GFlop/s
2.  $b_S$  = Applicable (saturated) peak bandwidth of the slowest data path utilized  
→ e.g.,  $b_S = 56$  GByte/s
3.  $I$  = Computational intensity (“work” per byte transferred) over the slowest data path utilized (code balance  $B_C = I^{-1}$ )  
→ e.g.,  $I = 0.167$  Flop/Byte →  $B_C = 6$  Byte/Flop

“Flop” is not the only  
useful unit of work!

Performance limit:

$$P = \min(P_{\max}, I \cdot b_S) = \min\left(P_{\max}, \frac{b_S}{B_C}\right)$$

[Byte/s] (pointing to  $b_S$ )

[Byte/Flop] (pointing to  $B_C$ )

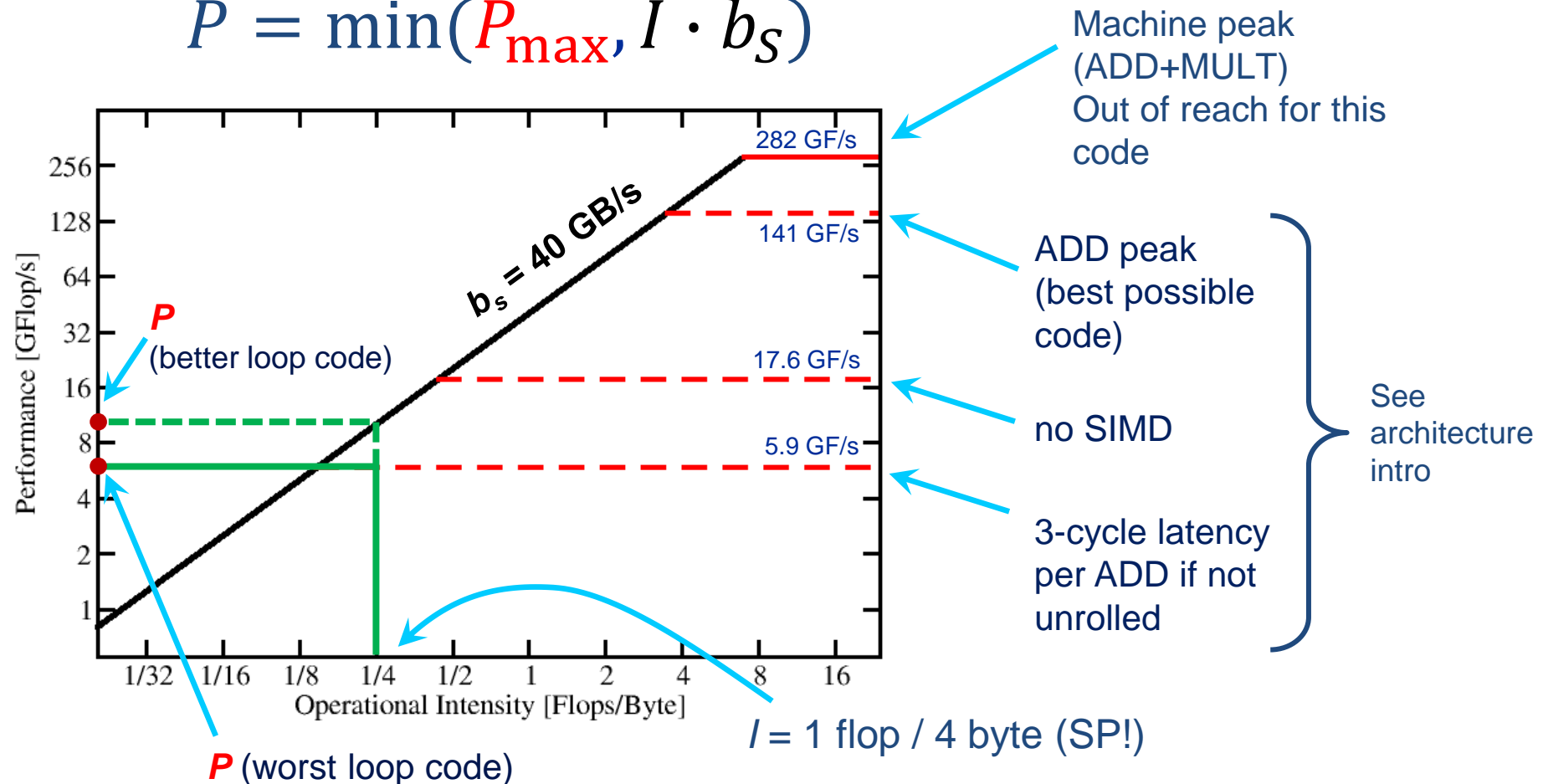


# Full Roofline for the sum reduction from the intro

Example: `do i=1,N; s=s+a(i); enddo`

in single precision on an 8-core 2.2 GHz Sandy Bridge socket @ “large” N

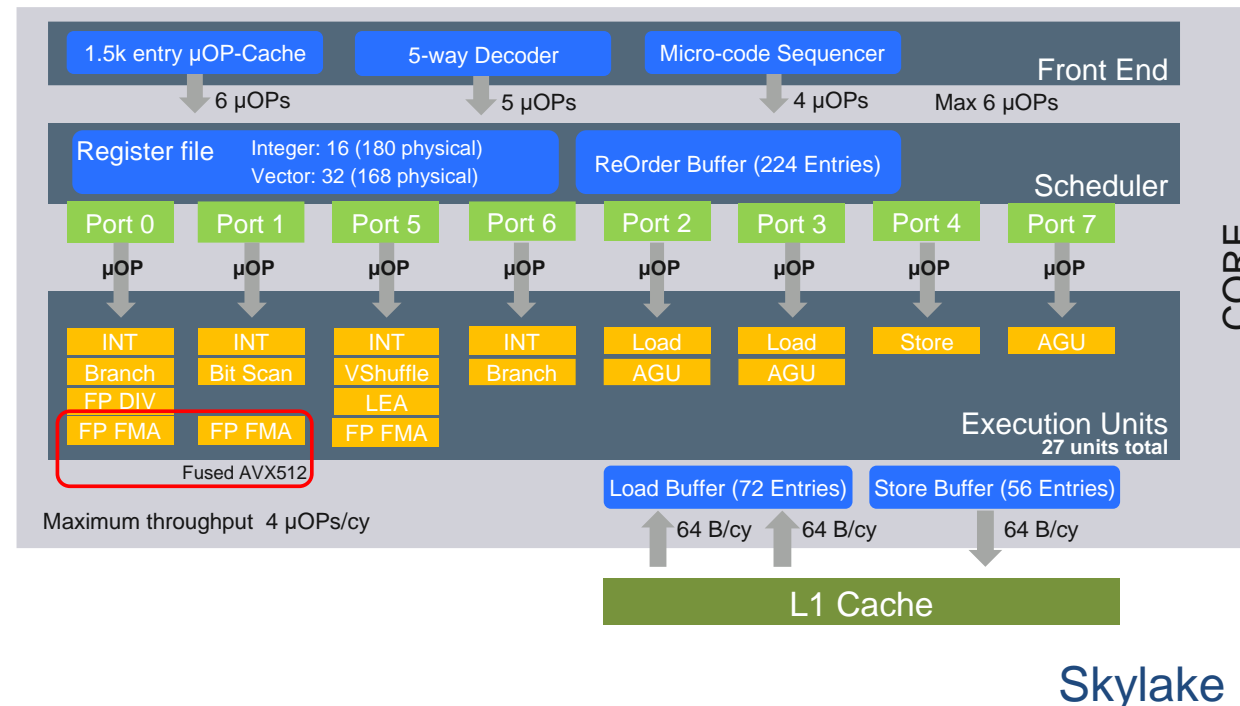
$$P = \min(P_{\max}, I \cdot b_s)$$



# Complexities of in-core execution ( $P_{\max}$ )

## Multiple bottlenecks:

- Decode/retirement throughput
- Port contention (direct or indirect)
- Arithmetic pipeline stalls (dependencies)
- Overall pipeline stalls (branching)
- L1 Dcache bandwidth (LD/ST throughput)
- Scalar vs. SIMD execution
- L1 Icache (LD/ST) bandwidth
- Alignment issues
- ...



Tool for  $P_{\max}$  analysis: OSACA

<http://tiny.cc/OSACA>

DOI: [10.1109/PMBS49563.2019.00006](https://doi.org/10.1109/PMBS49563.2019.00006)

DOI: [10.1109/PMBS.2018.8641578](https://doi.org/10.1109/PMBS.2018.8641578)

# Hardware features of (some) Intel Xeon processors

Microarchitecture	Ivy Bridge EP	Broadwell EP	Cascade Lake SP	Ice Lake SP
Introduced	09/2013	03/2016	04/2019	06/2021
Cores	≤ 12	≤ 22	≤ 28	≤ 40
LD/ST throughput per cy:				
AVX(2), AVX512	1 LD + ½ ST	2 LD + 1 ST	2 LD + 1 ST	2 LD + 1 ST
SSE/scalar	2 LD    1 LD & 1 ST			
ADD throughput	1 / cy	1 / cy	2 / cy	2 / cy
MUL throughput	1 / cy	2 / cy	2 / cy	2 / cy
FMA throughput	N/A	2 / cy	2 / cy	2 / cy
L1-L2 data bus	32 B/cy	64 B/cy	64 B/cy	64 B/cy
L2-L3 data bus	32 B/cy	32 B/cy	16+16 B/cy	16+16 B/cy
L1/L2 per core	32 KiB / 256 KiB	32 KiB / 256 KiB	32 KiB / 1 MiB	48 KiB / 1.25 MiB
LLC	2.5 MiB/core inclusive	2.5 MiB/core inclusive	1.375 MiB/core exclusive/victim	1.5 MiB/core exclusive/victim
Memory	4ch DDR3	4ch DDR3	6ch DDR4	8ch DDR4
Memory BW (meas.)	~ 48 GB/s	~ 62 GB/s	~ 115 GB/s	~ 160 GB/s

Source:

<https://software.intel.com/content/www/us/en/develop/download/intel-64-and-ia-32-architectures-optimization-reference-manual.html>

# Code balance: more examples

```
double a[], b[];
for(i=0; i<N; ++i)
    a[i] = a[i] + b[i];
```

$$B_C = 24B / 1F = 24 \text{ B/F}$$
$$I = 0.042 \text{ F/B}$$

```
double a[], b[];
for(i=0; i<N; ++i)
    a[i] = a[i] + s * b[i];
```

$$B_C = 24B / 2F = 12 \text{ B/F}$$
$$I = 0.083 \text{ F/B}$$

```
float s=0, a[];
for(i=0; i<N; ++i)
    s = s + a[i] * a[i];
```

Scalar – can be kept in register

$$B_C = 4B / 2F = 2 \text{ B/F}$$
$$I = 0.5 \text{ F/B}$$

```
float s=0, a[], b[];
for(i=0; i<N; ++i)
    s = s + a[i] * b[i];
```

Scalar – can be kept in register

$$B_C = 8B / 2F = 4 \text{ B/F}$$
$$I = 0.25 \text{ F/B}$$

```
float s=0, a[], b[];
for(i=0; i<N; ++i)
    for(j=0; j<N; ++j)
        b[i][j] = a[i][j]
                + a[i-1][j]
                + a[i+1][j];
```

$$B_C = 16B / 2F \text{ or}$$
$$8B / 2F \text{ or even } ???$$
$$20 \text{ B} / 2F$$

Streaming, perfect spatial locality, no temporal locality  
→ simple

And what about this?

```
float s=0, a[], b[];
int idx[];
for(i=0; i<N; ++i)
    s = s + a[i]
        * b[idx[i]];
```

Possible cache reuse → tricky!

We'll get to it!

# Is there anything to ease the construction of the model?

## Code balance $B_C$

- Close inspection and hard thinking
- Simplifying assumptions
  - “What is the minimum possible amount of traffic?”
  - “What is the worst case?”
- Tools
  - Kerncraft  
<https://github.com/RRZE-HPC/kerncraft>

## In-core $P_{\max}$

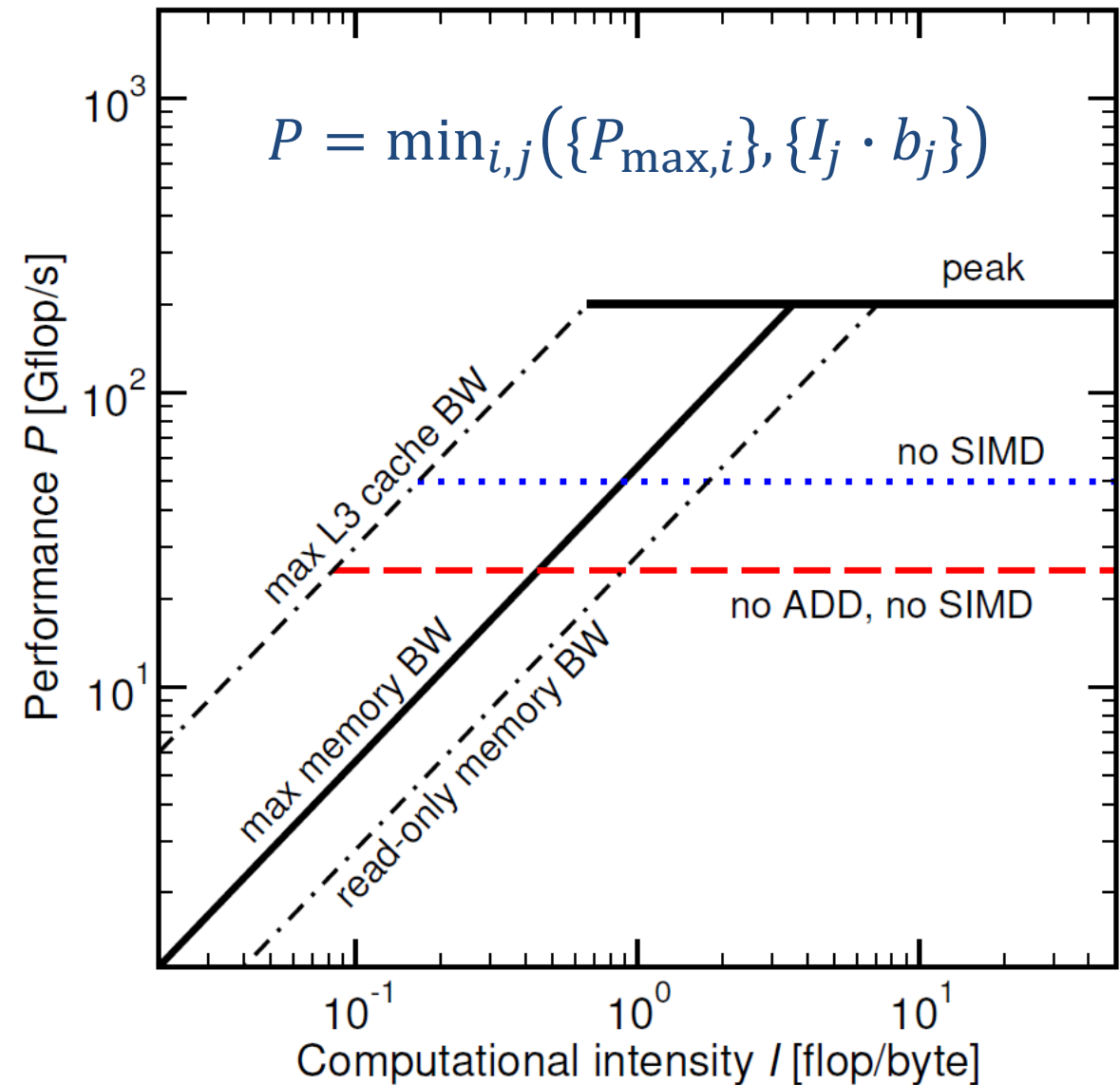
- Inspection of assembly code and manual modeling
- Simplifying assumptions
  - “What is the required minimum number of arithmetic/load/store instructions?”
  - $P_{\max} = P_{peak}$
- Tools
  - OSACA  
<https://github.com/RRZE-HPC/OSACA>

# Refined Roofline model: graphical representation

## Multiple ceilings may apply

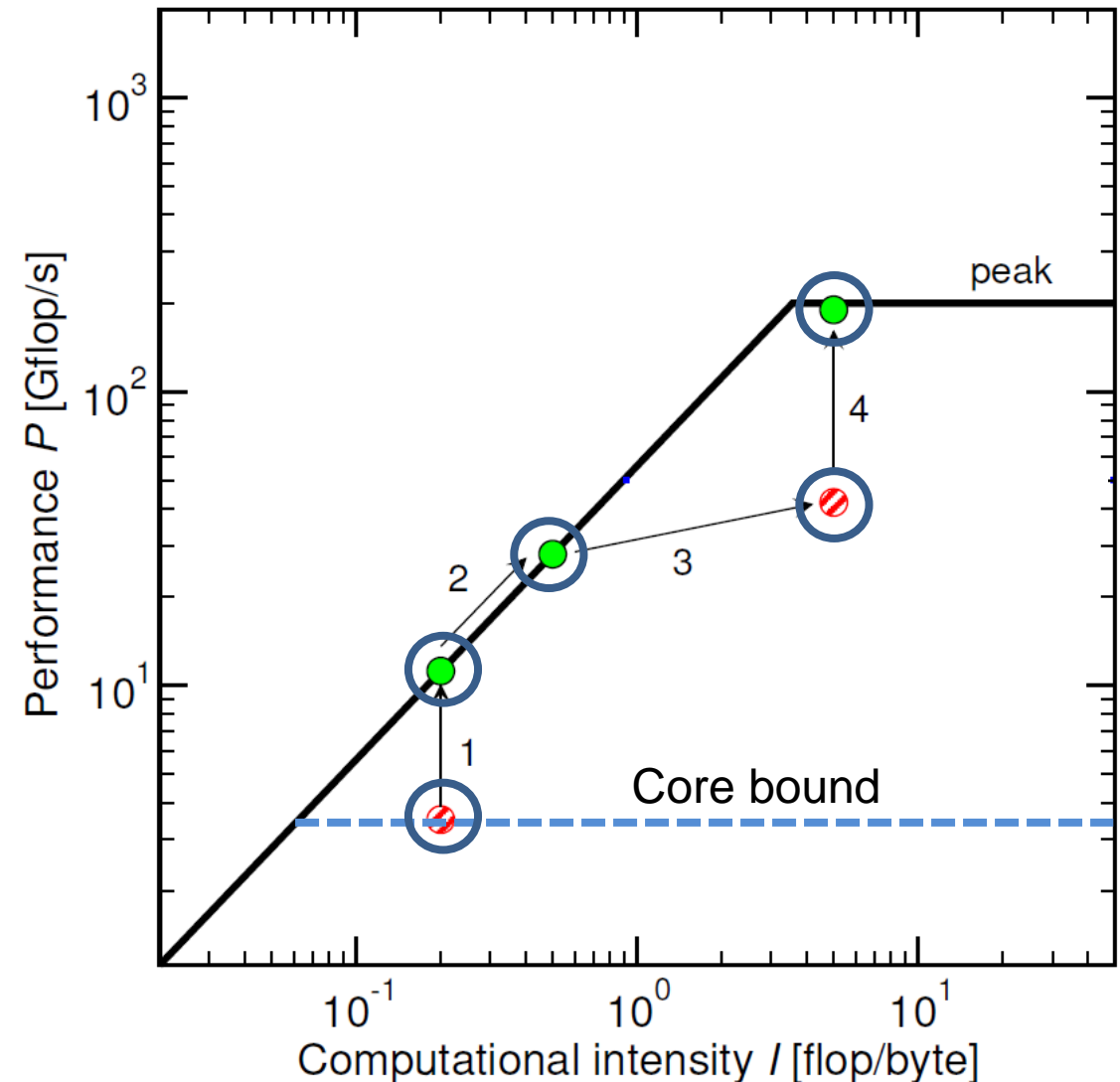
- Different **bandwidths / data paths**  
→ different inclined ceilings  
→ possibly different  $I$  for one kernel
- Different  $P_{\max}$   
→ different flat ceilings

In fact,  $P_{\max}$  should always come from **code analysis**; generic ceilings are usually impossible to attain



# Tracking code optimizations in the Roofline Model

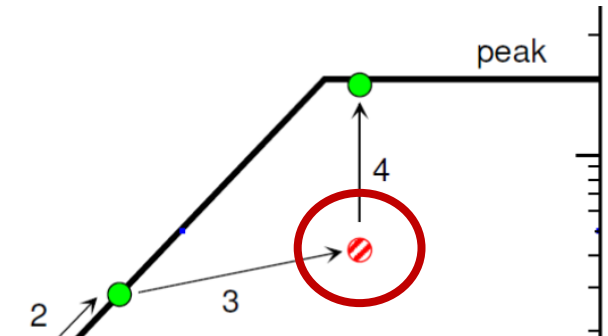
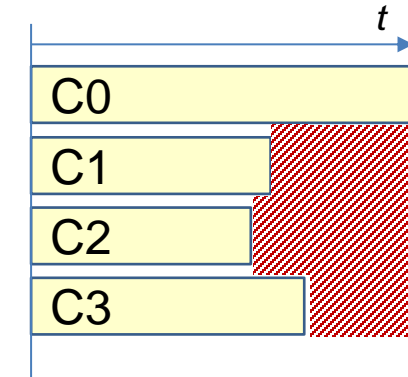
1. Hit the BW bottleneck by good serial code  
(e.g., Ninja C++ → Fortran)
2. Increase intensity to make better use of BW bottleneck  
(e.g., spatial loop blocking)
3. Increase intensity and go from memory bound to core bound  
(e.g., temporal blocking)
4. Hit the core bottleneck by good serial code  
(e.g., `-fno-alias`, SIMD intrinsics)



# Roofline: How can it “fail”?

... assuming that you did the math right?

- **Load imbalance**
  - May be impossible to saturate memory bandwidth
  - This includes serial code
- **“Slow code”**
  - “Invisible” performance ceiling due to **inefficient instructions** or **inefficient execution**
- **Erratic memory access patterns**
  - **Latency** rains on your parade



```
for(int i=0; i<N; ++i)
    a[i] = s * b[index[i]];
```

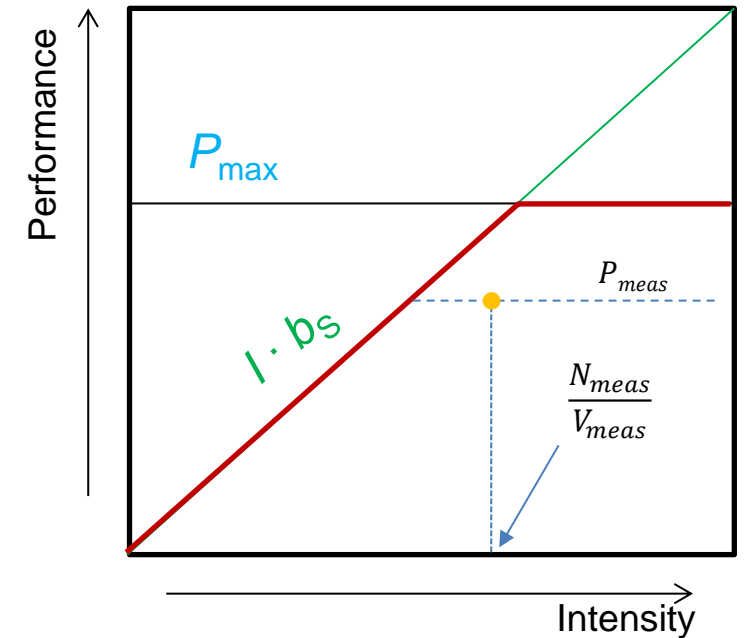


# Diagnostic / phenomenological Roofline modeling



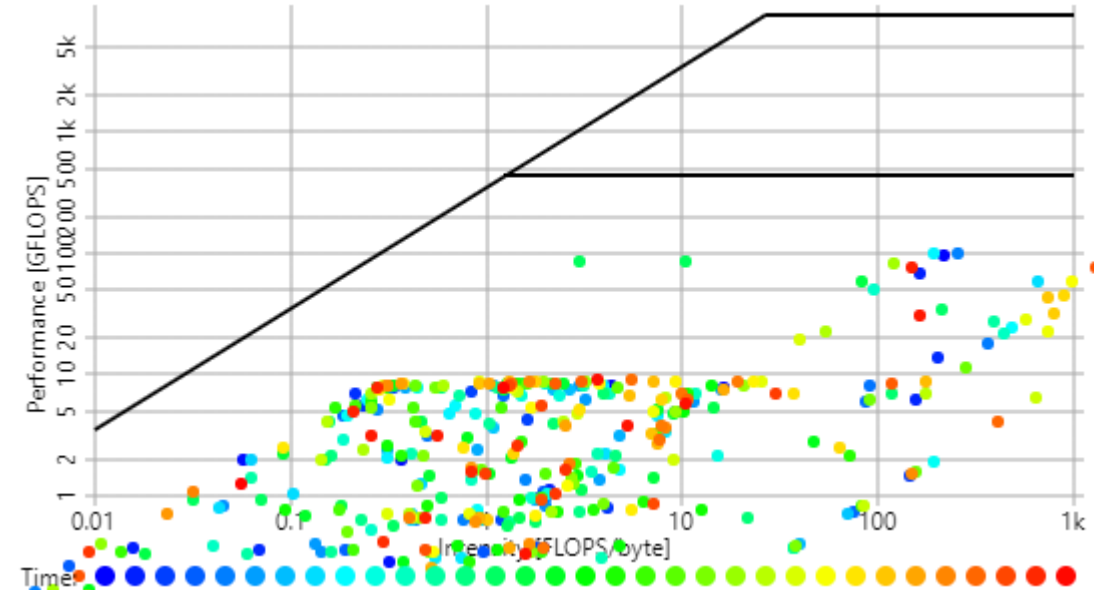
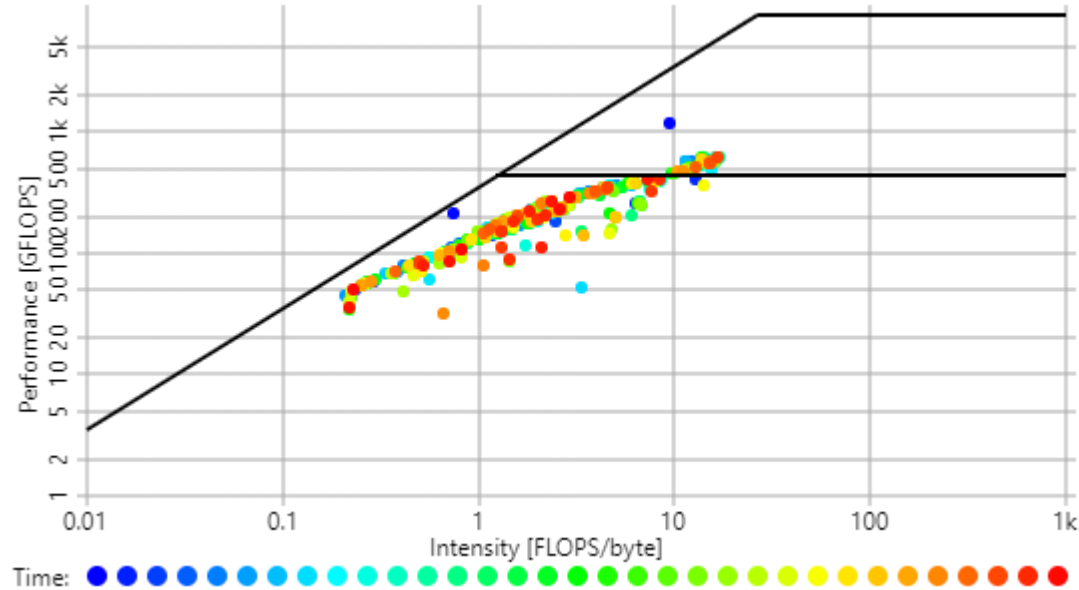
# Diagnostic modeling

- What if we cannot predict the intensity/balance?
  - Code very complicated
  - Code not available
  - Parameters unknown
  - Doubts about correctness of analysis
- Measure data volume  $V_{meas}$  (and work  $N_{meas}$ )
  - Hardware performance counters
  - Tools: likwid-perfctr, PAPI, Intel Vtune,...
- Insights + benefits
  - Compare analytic model and measurement → validate model
  - Can be applied (semi-)automatically
  - Useful in performance monitoring of user jobs on clusters



# Roofline and performance monitoring of clusters

Two cluster jobs...



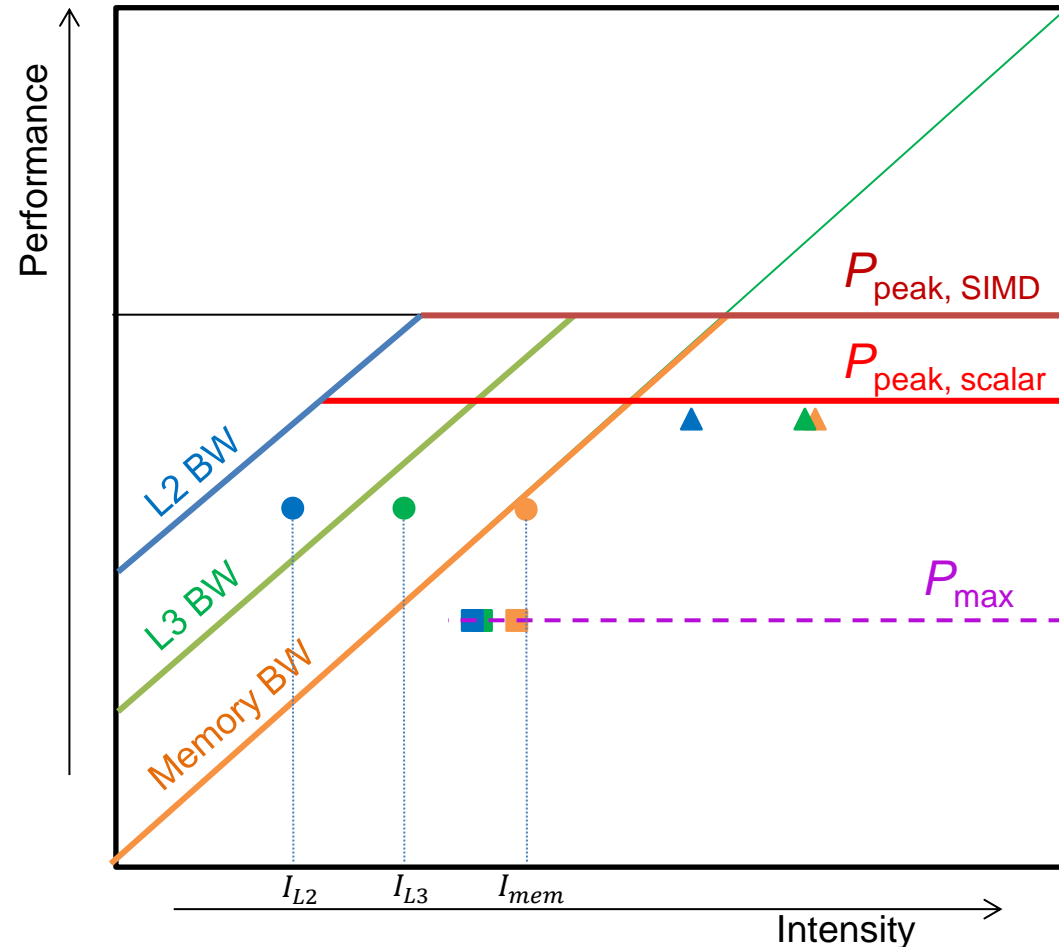
Which of them is  
“good” and which is  
“bad”?



# Diagnostic modeling of a complex code (3 kernels)

Multiple bandwidth bottlenecks

→ need  $I$  for each one ( $I_{mem}, I_{L3}, I_{L2}, \dots$ )



## Kernel 1 ●

- Performance close to memory BW ceiling but far away from others  
→ indicates **memory bound**

## Kernel 2 ▲

- Performance not near any BW ceiling
- Performance close to scalar peak ceiling  
→ indicates **scalar core-bound peak code**

## Kernel 3 ■

- Performance not anywhere near any ceiling  
→ There must be an (as yet) **unknown in-core performance limit**  $P_{max}$

# Roofline conclusion

- Roofline = simple first-principle model for upper performance limit of data-streaming loops
  - Machine model ( $P_{max}, b_S, \dots$ ) + application model ( $I_{mem}, \dots$ )
  - Conditions apply, extensions exist
- Two modes of operation; per kernel:
  - Predictive: Calculate  $I_j$ , calculate upper limit, validate model, optimize, iterate
  - Diagnostic: Measure  $I_j$  and  $P$ , compare with ceilings
- Challenge of predictive modeling: Getting  $P_{max}$  and  $I$  right