



# Modern computer architecture

#### An introduction for software developers

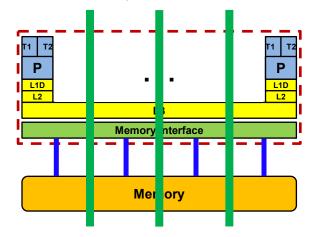


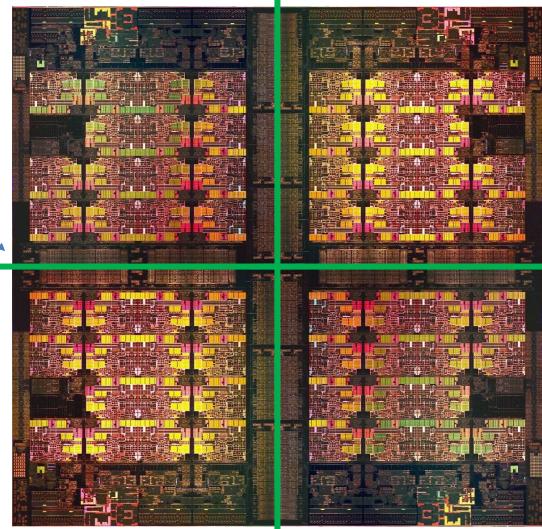
## Multi-core today: Intel Xeon Sapphire Rapids (2023)

- Xeon "Sapphire Rapids" (Platinum/Gold/Silver/Bronze): Up to 60 cores running at 1.7+ GHz (+ "Turbo Mode" 4.8 GHz),
- Simultaneous Multithreading
   → reports as 120-way chip
- "Intel 7" process / up to 350 W
- Multi-die package (4 chips)
- Clock frequency: flexible <sup>(i)</sup>

Optional: "Sub-NUMA Clustering" (SNC) mode boot option

→ One memory domain per die



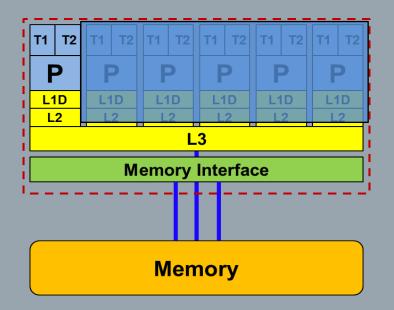


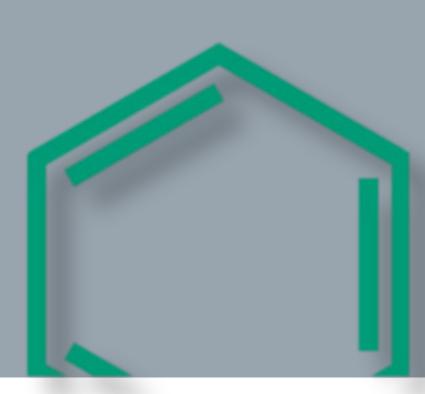
https://www.techpowerup.com/292204/intel-sapphire-rapids-xeon-4-tile-mcm-annotated



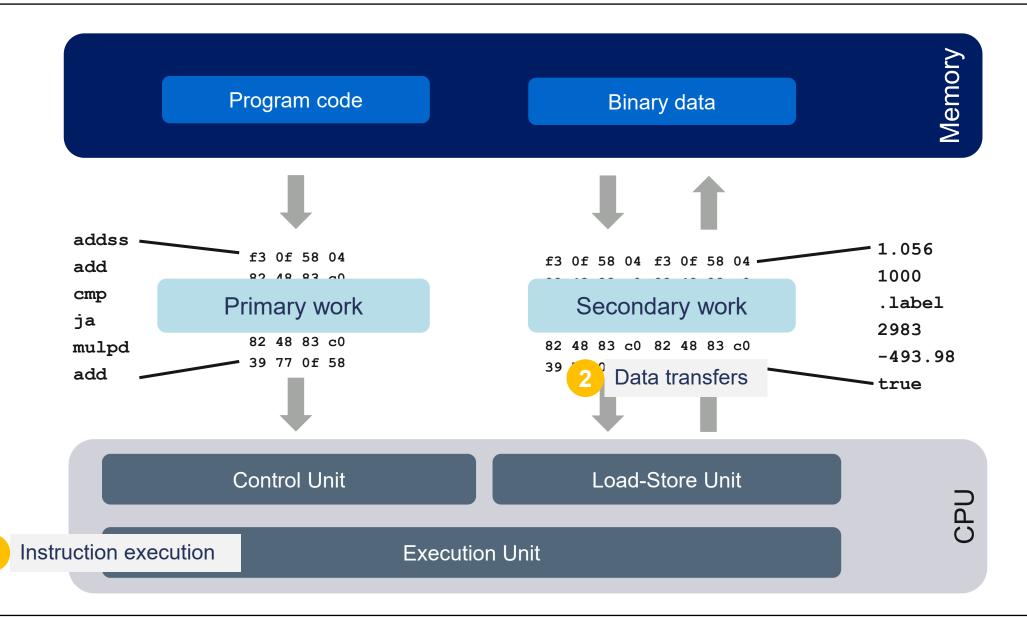


# A deeper dive into core architecture

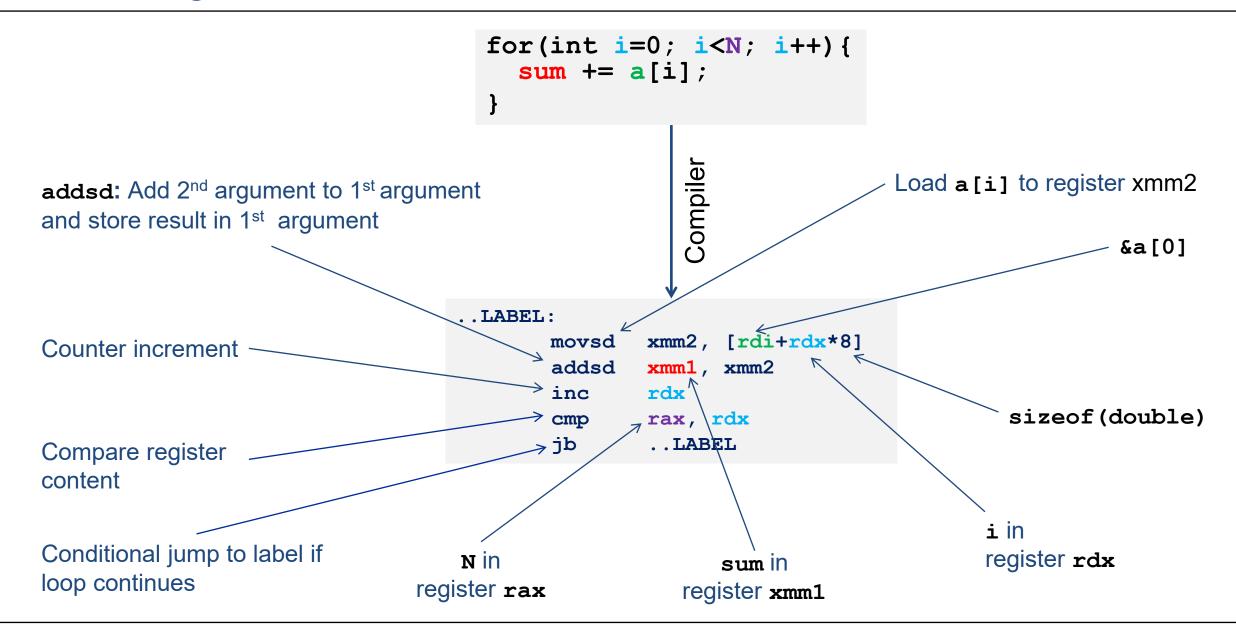




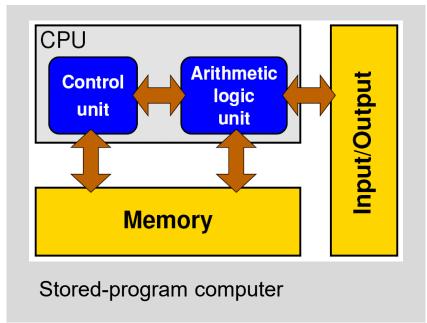
#### **Stored Program Computer**



#### From high level code to actual execution

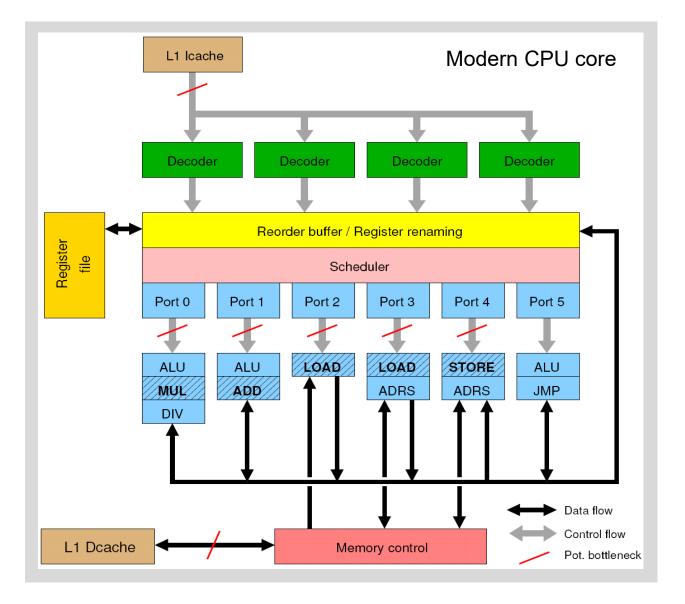


#### General-purpose cache based microprocessor core



- Implements "Stored Program Computer" concept (Turing 1936)
- Similar designs on all modern systems
- (Still) multiple potential bottlenecks

The clock cycle is the "heartbeat" of the core

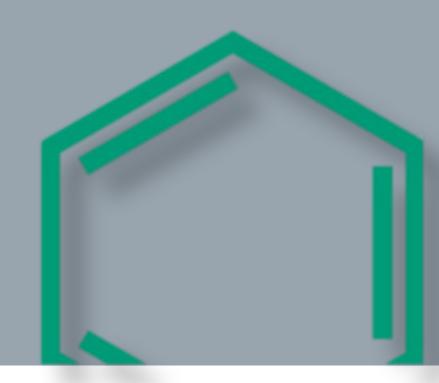




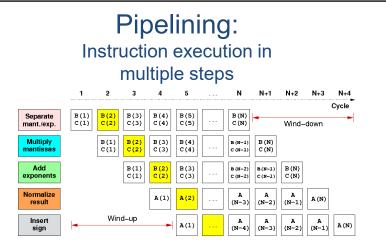


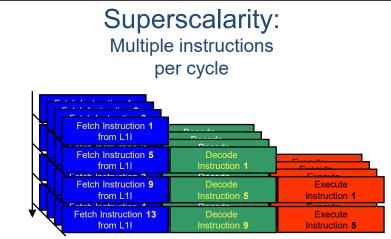
# **In-core features**

#### Pipelining, Superscalarity, SIMD, SMT



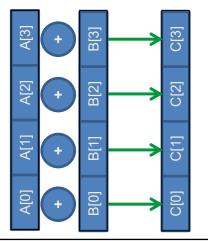
#### Important in-core features



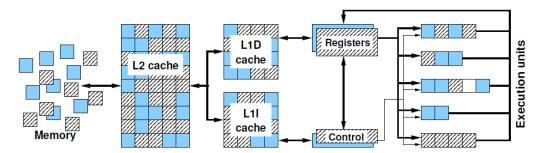


#### Single Instruction Multiple Data:

Multiple operations per instruction



#### Simultaneous Multi-Threading: Multiple instruction sequences in parallel



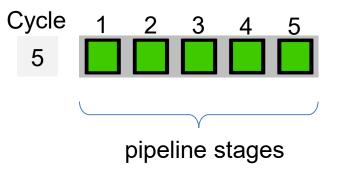
# Instruction level parallelism (ILP): pipelining, superscalarity

#### Pipelining

Independent instructions (of one kind, e.g., ADD):

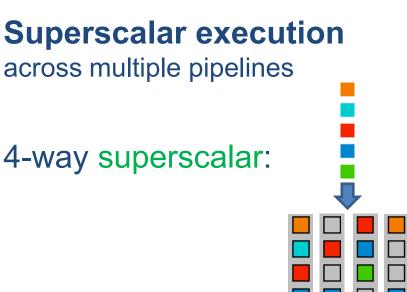
15	14	13	12	11

Single instruction takes 5 cycles (latency)



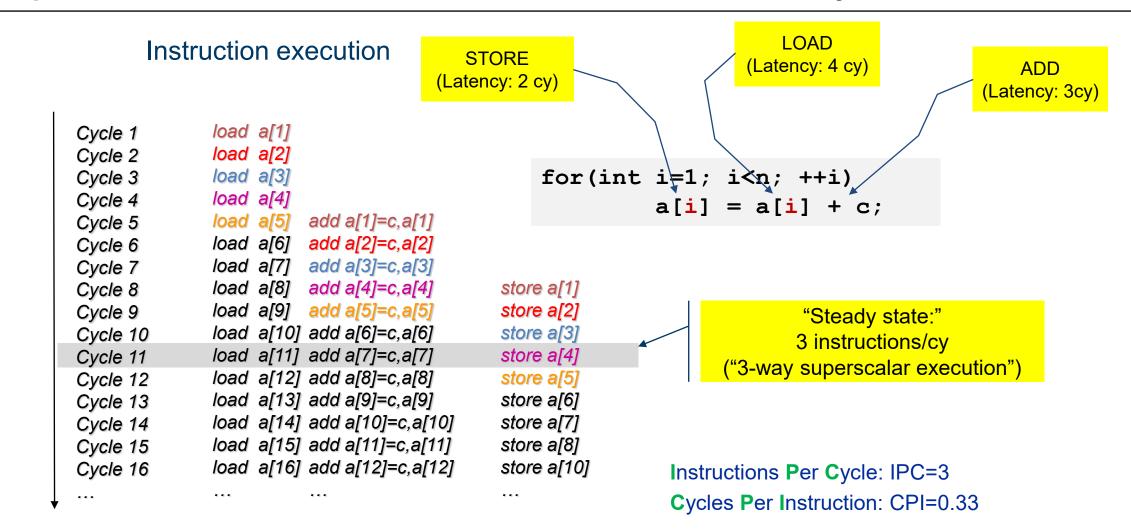
#### Throughput:

- 1 instruction per cycle after pipeline is full
- $\rightarrow$  5x speedup



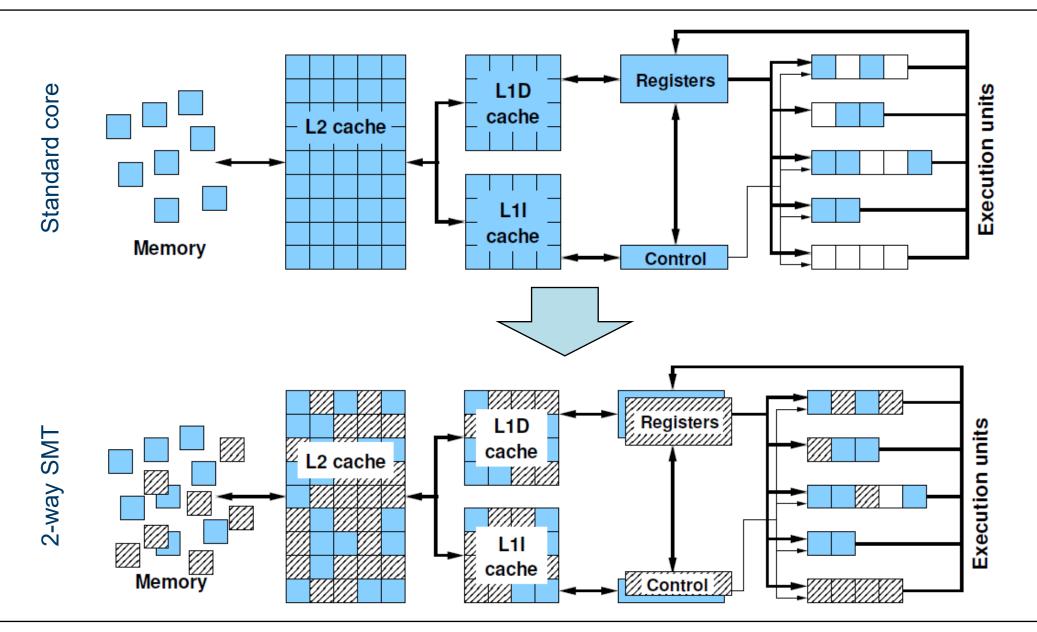
- → Massive boost in instruction throughput
- → Instructions can be reordered on the fly

#### Superscalar out-of-order execution and steady state



Hardware takes care of executing instructions as soon as their operands are available: Out-Of-Order (OOO) execution

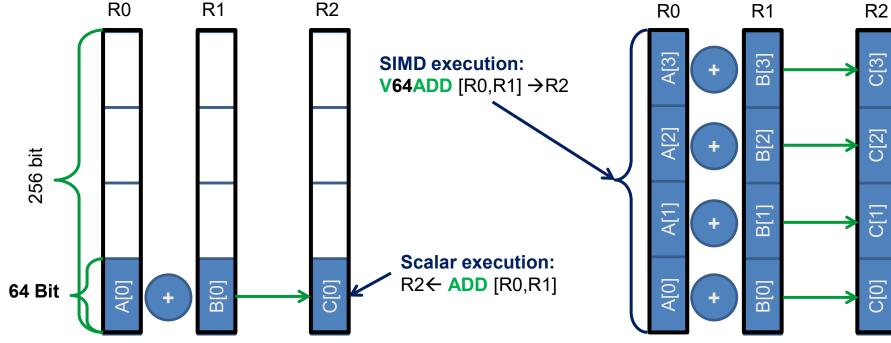
## Simultaneous multi-threading (SMT)



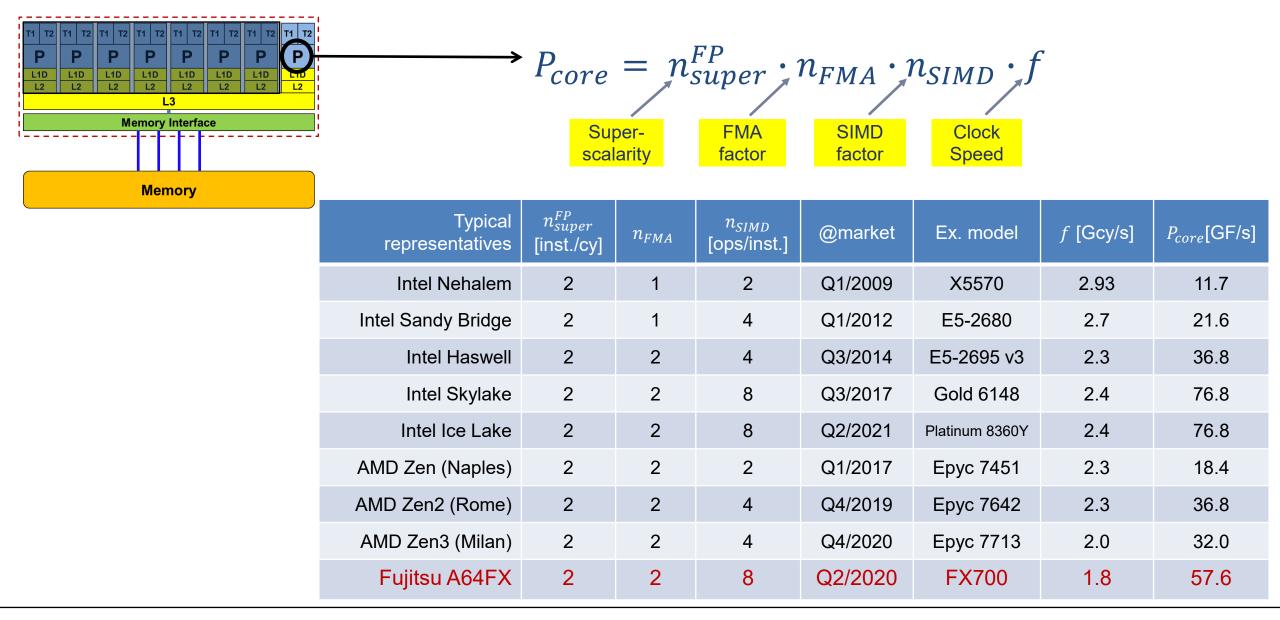
#### **Basic Node Architecture**

## SIMD processing

- Single Instruction Multiple Data (SIMD) operations allow the execution of the same operation on "wide" registers from a single instruction
- x86 SIMD instruction sets:
  - SSE: register width = 128 Bit  $\rightarrow$  2 double precision floating point operands
  - AVX: register width = 256 Bit  $\rightarrow$  4 double precision floating point operands
  - AVX-512: ... you guessed it!
- Adding two registers holding double precision floating point operands:

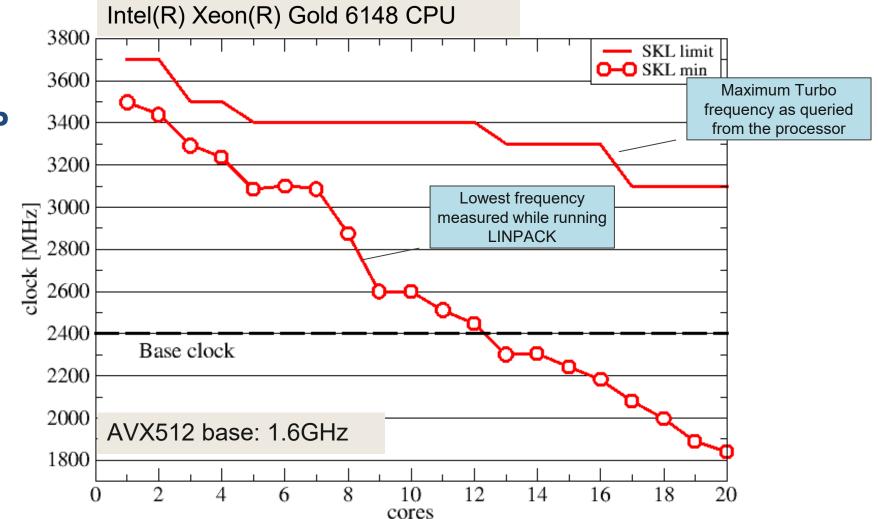


## Single-core DP floating-point performance



## Multi-core today: Turbo mode

The processor dynamically overclocks to exploit more of the **TDP** envelope if fewer cores are active







# **Example: The sum reduction**



```
for (int i=0; i<N; i++) {
    sum += a[i];
}</pre>
```

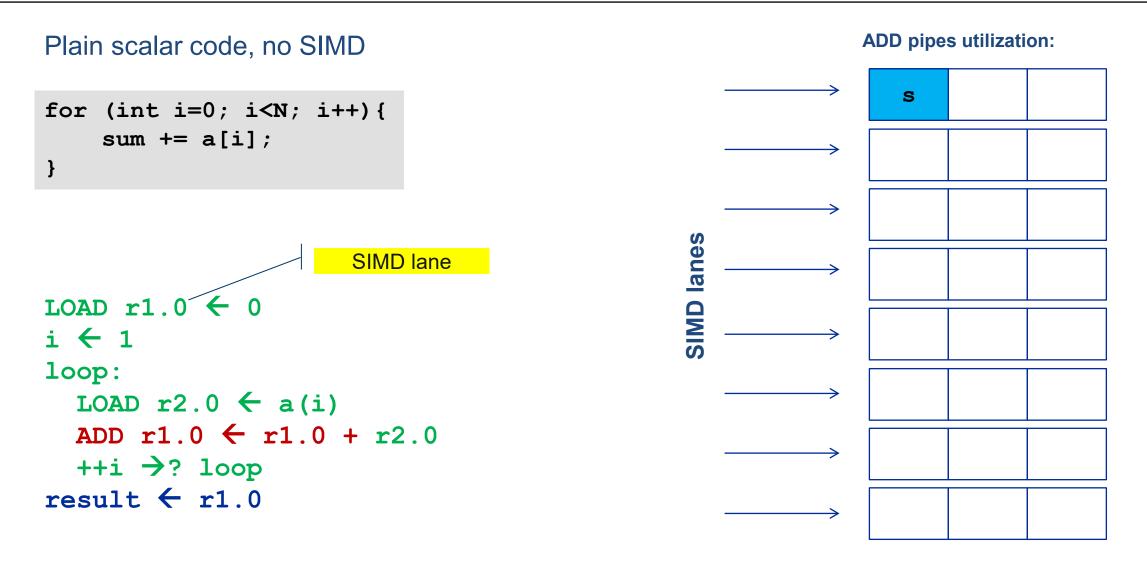
...In single precision on an AVXcapable core (ADD latency = 3 cy)

How fast can this loop possibly run with data in the L1 cache?

- Loop-carried dependency on summation variable
- Execution stalls at every ADD until previous ADD is complete

→No pipelining?→No SIMD?

## Applicable peak for the sum reduction (I)

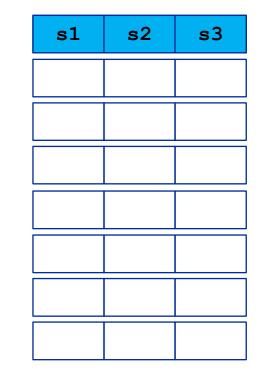


 $\rightarrow$  1/24 of ADD peak

## Applicable peak for the sum reduction (II)

Scalar code, 3-way "modulo variable expansion" LOAD r1.0  $\leftarrow$  0 LOAD  $r2.0 \leftarrow 0$ LOAD r3.0  $\leftarrow$  0 i ← 1 loop: LOAD r4.0  $\leftarrow$  a(i) LOAD r5.0  $\leftarrow$  a(i+1) LOAD r6.0  $\leftarrow$  a(i+2) ADD  $r1.0 \leftarrow r1.0 + r4.0 \# scalar ADD$ ADD  $r2.0 \leftarrow r2.0 + r5.0 \# scalar ADD$ ADD  $r3.0 \leftarrow r3.0 + r6.0 \# scalar ADD$ i+=3 **→**? loop result  $\leftarrow$  r1.0+r2.0+r3.0

```
for (int i=0; i<N; i+=3) {
    s1 += a[i+0];
    s2 += a[i+1];
    s3 += a[i+2];
}
sum = sum + s1+s2+s3;</pre>
```



 $\rightarrow$  1/8 of ADD peak

## Applicable peak for the sum reduction (III)

```
SIMD vectorization (8-way MVE) x
      pipelining (3-way MVE)
```

```
LOAD [r1.0, ..., r1.7] \leftarrow [0, ..., 0]
LOAD [r2.0, ..., r2.7] \leftarrow [0, ..., 0]
LOAD [r3.0, ..., r3.7] \leftarrow [0, ..., 0]
i ← 1
```

result  $\leftarrow$  r1.0+r1.1+...+r3.6+r3.7

```
for (int i=0; i<N; i+=24) {</pre>
  s10 += a[i+0]; s20 += a[i+8]; s30 += a[i+16];
 s11 += a[i+1]; s21 += a[i+9]; s31 += a[i+17];
 s12 += a[i+2]; s22 += a[i+10]; s32 += a[i+18];
 s13 += a[i+3]; s23 += a[i+11]; s33 += a[i+19];
 s14 += a[i+4]; s24 += a[i+12]; s34 += a[i+20];
 s15 += a[i+5]; s25 += a[i+13]; s35 += a[i+21];
 s16 += a[i+6]; s26 += a[i+14]; s36 += a[i+22];
 s17 += a[i+7]; s27 += a[i+15]; s37 += a[i+23];
```

```
sum = sum + s10 + s11 + ... + s37;
```

		s10	s20	s30
oop:		s11	s21	s31
LOAD $[r4.0,,r4.7] \leftarrow [a(i),,a(i+7)] \# SIMD LOAD$ LOAD $[r5.0,,r5.7] \leftarrow [a(i+8),,a(i+15)] \# SIMD$	~	s12	s22	s32
LOAD [r6.0,,r6.7] < [a(i+16),,a(i+23)] # SIMD	peak	s13	s23	s33
ADD r1 $\leftarrow$ r1 + r4 # SIMD ADD	ADD	s14	s24	s34
ADD $r2 \leftarrow r2 + r5 \# SIMD ADD$	1	s15	s25	s35
ADD r3 $\leftarrow$ r3 + r6 # SIMD ADD		s16	s26	s36
$i + = 24 \rightarrow ?$ loop esult $\leftarrow$ r1 0+r1 1+ +r3 6+r3 7		s17	s27	s37

loop:

# Sum reduction

#### Questions

- When can this performance actually be achieved?
  - No data transfer bottlenecks
  - No other in-core bottlenecks
    - Need to execute (3 LOADs + 3 ADDs + 1 increment + 1 compare + 1 branch) in 3 cycles
- What does the compiler do?
  - If allowed and capable, the compiler will do this automatically
- Is the compiler allowed to do this at all?
  - Not according to language standards
  - High optimization levels can violate language standards
- What about the "accuracy" of the result?
  - Good question ;-)



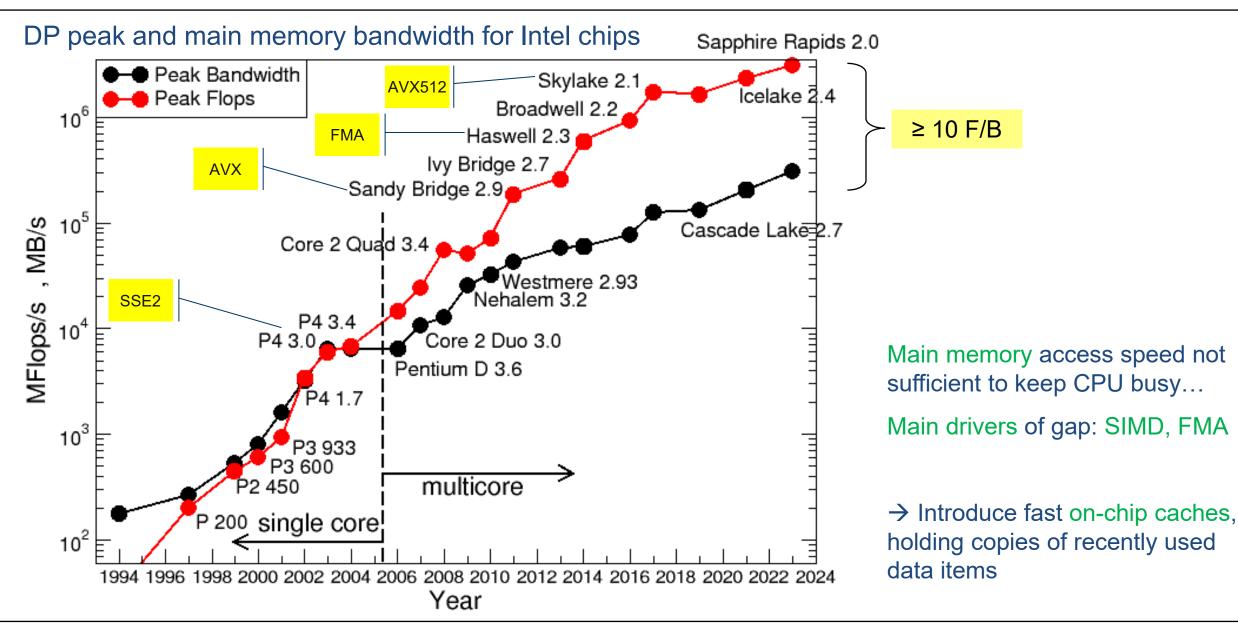


# **Memory Hierarchy**

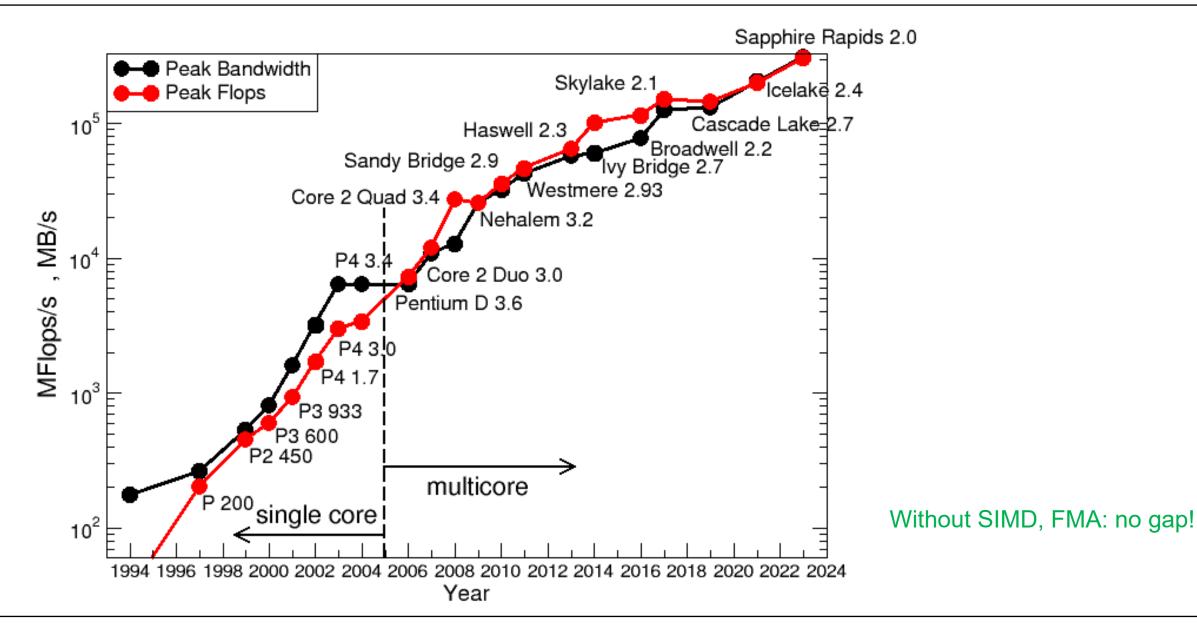
#### In-cache performance (L2, L3) Main memory performance



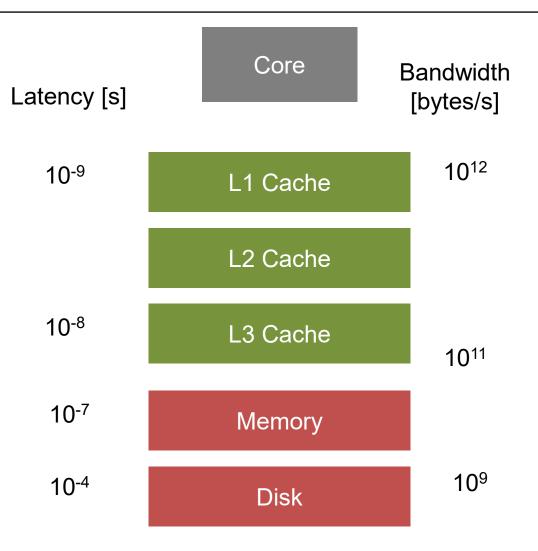
## Von Neumann bottleneck reloaded: "DRAM gap"



# The "stripped" von Neumann bottleneck



You can either build a small and fast memory or a large and slow memory

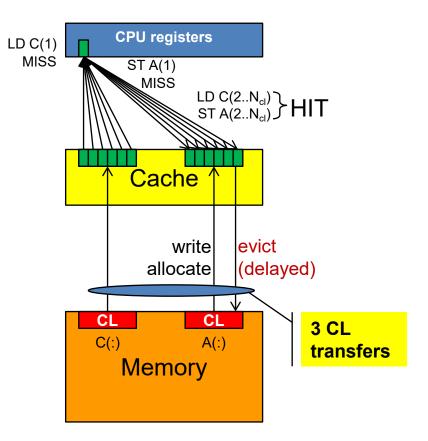


#### Purpose of many optimizations: use data in fast memory

## Data transfers in a memory hierarchy

Caches help with getting instructions and data to the CPU "fast" How does data travel from memory to the CPU and back?

- Remember: Caches are organized in cache lines (e.g., 64 bytes)
- Only complete cache lines are transferred between memory hierarchy levels (except registers)
- Registers can only "talk" to the L1 cache
- MISS: Load or store instruction does not find the data in acache level
  - $\rightarrow$  CL transfer required



Example: Array copy A(:)=C(:)

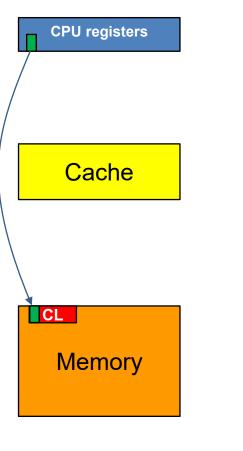
## Avoiding the write-allocate transfer

#### Disadvantages of write-allocate:

- Cache pollution (if data not needed anytime soon)
- Additional data traffic

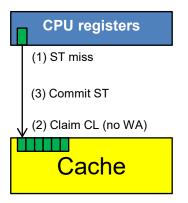
#### Solution 1: Nontemporal stores

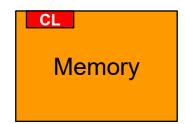
- A.k.a. "streaming stores," store instruction with a "nontemporal hint"
- Write "directly" to memory, ignoring the normal cache hierarchy
- Avoids cache pollution, but stored data ends up in memory



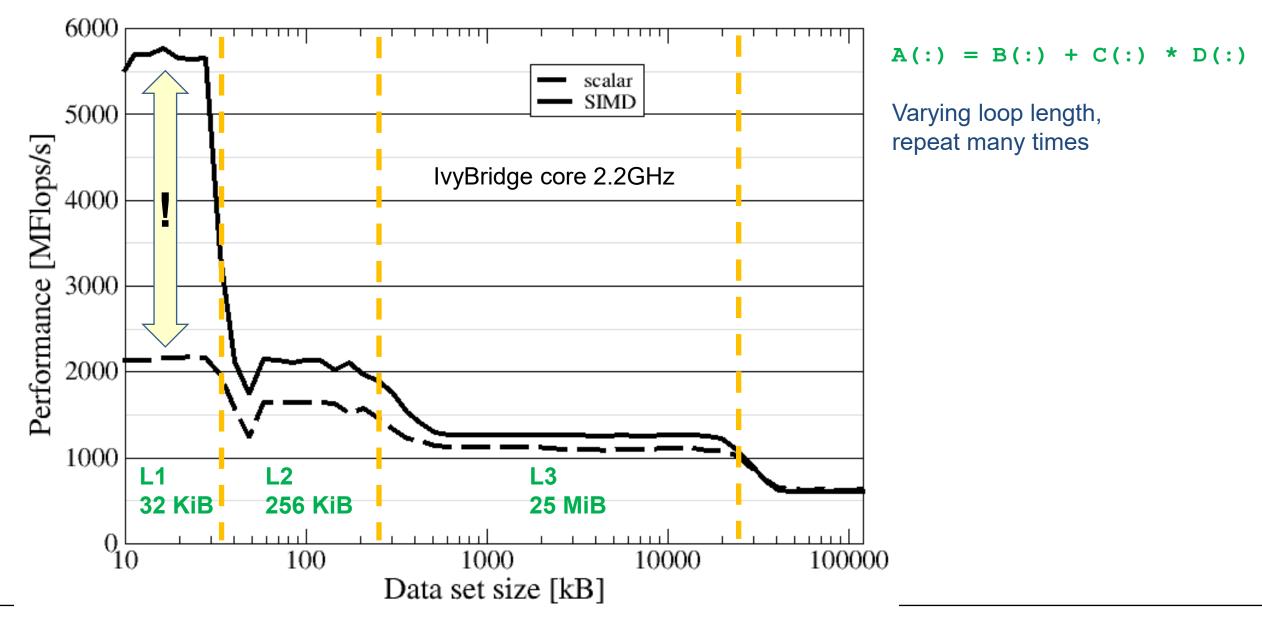
Solution 2: Cache line claim

- Special instructions (e.g., on POWER, A64FX) or automatic in hardware (Arm, Intel Ice Lake)
- Core claims CL in some level when guranteed to be overwritten completely
- Allows stored data to remain in cache
   → does not reduce cache pollution





## Getting the data from far away





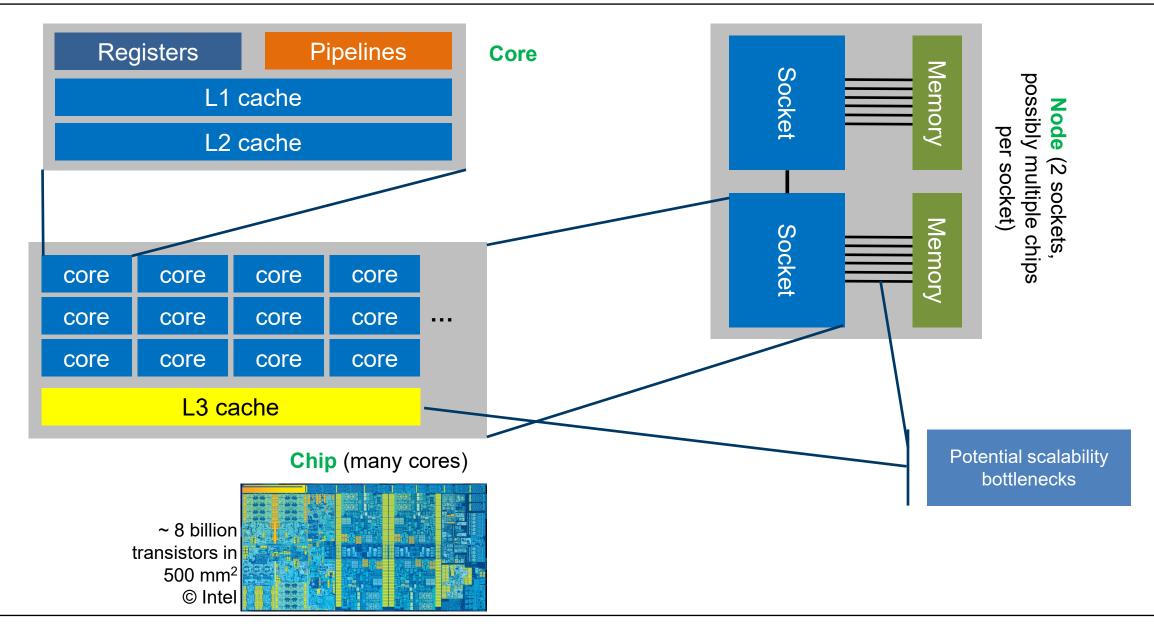


# **Multicore Chips**

Memory bandwidth scaling Node topology and performance



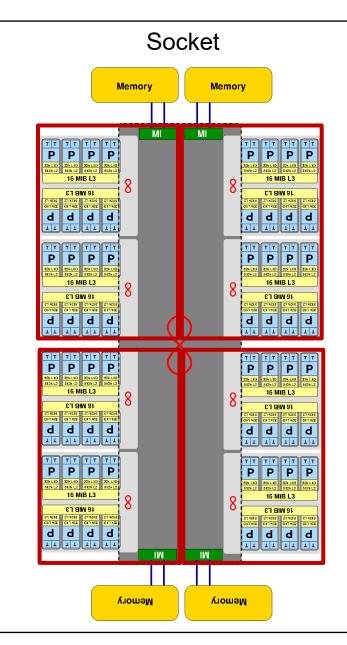
#### Node topology of HPC systems



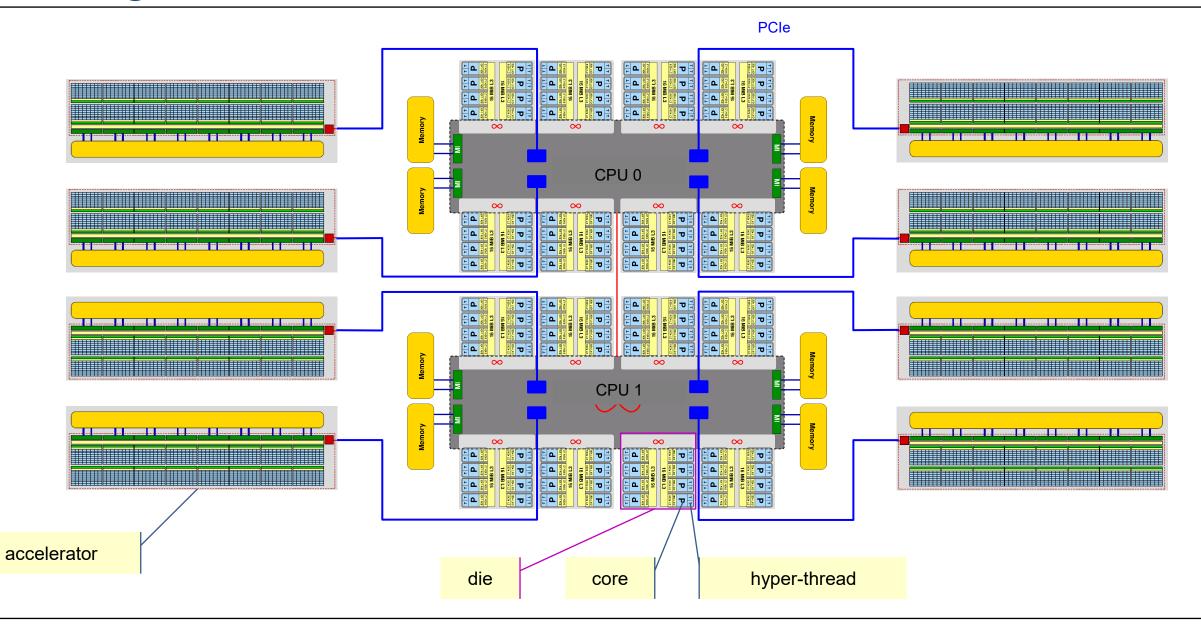
Basic Node Architecture

#### Putting the cores & caches together AMD Epyc 7742 64-Core Processor («Rome»)

- Core features:
  - Two-way SMT
  - Two 256-bit SIMD FMA units (AVX2)
     →16 flops/cycle
  - 32 KiB L1 data cache per core
  - 512 KiB L2 cache per core
- 64 cores per socket hierarchically built up from
  - I6 CCX with 4 cores and 16 MiB of L3 cache
  - 2 CCX form 1 CCD (silicon die)
  - 8 CCDs connected to IO device "Infinity Fabric" (memory controller & PCIe)
- 8 channels of DDR4-3200 per IO device
  - MemBW: 8 ch x 8 byte x 3.2 GHz = 204.8 GB/s
- ccNUMA feature (boot time option):
  - Nodes Per Socket (NPS)=1, 2 or 4
  - NPS=4  $\rightarrow$  4 ccNUMA domains



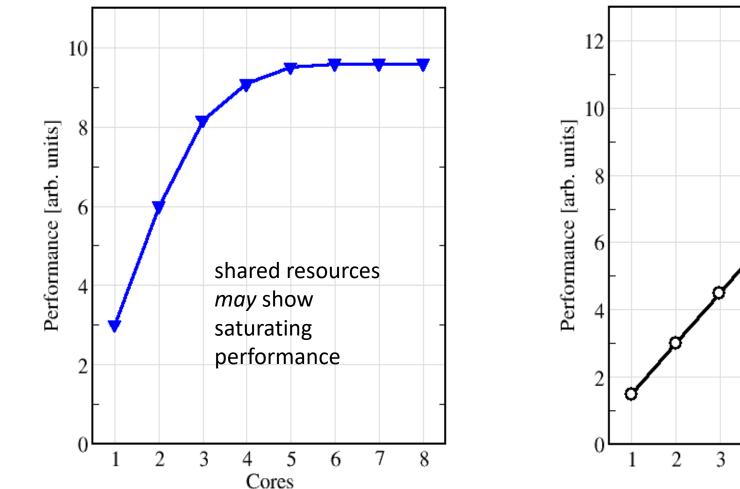
# Adding accelerators to the node

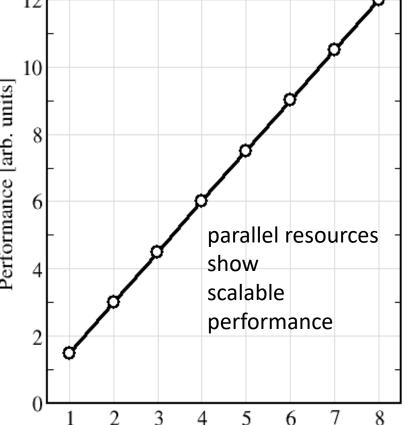


**Basic Node Architecture** 

#### Scalable and saturating behavior

Clearly distinguish between "**saturating**" and "**scalable**" performance on the chip level One of the most important performance signatures

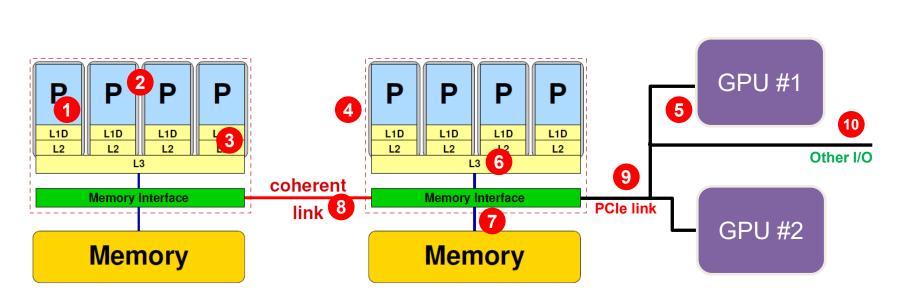




Cores

## Parallelism in a modern compute node

Parallel and shared resources within a shared-memory node



#### **Parallel resources:**

- Execution/SIMD units 1
- Cores
- Inner cache levels 3
- Sockets / ccNUMA domains
- Multiple accelerators 5

#### **Shared resources:**

- Outer cache level per socket
- Memory bus per socket 7
- Intersocket link 8
- PCIe bus(es) 9
- Other I/O resources 10

#### How does your application react to all of those details?

4





# Interlude: A glance at accelerator technology

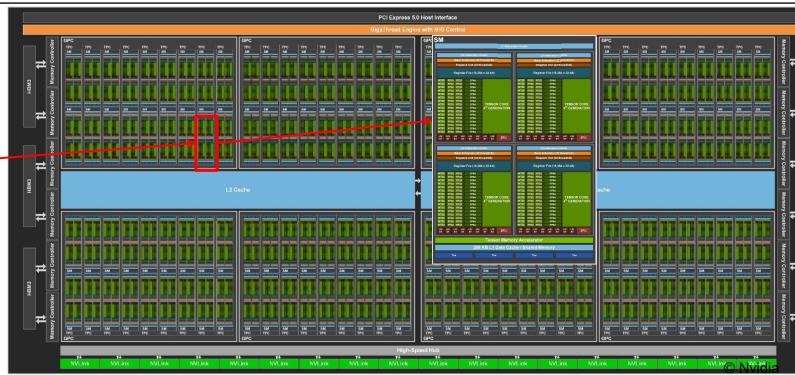
NVIDIA "Hopper" H100 vs. AMD Zen4 "Genoa"



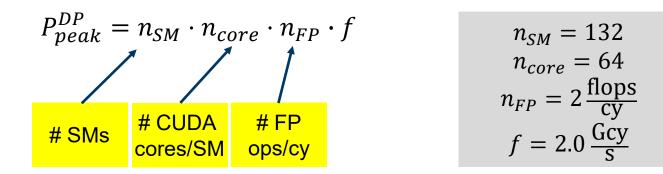
# Nvidia H100 "Hopper" SXM5 (700 W) specs

#### Architecture

- 80 B Transistors
- ~ 2.0 GHz clock speed (turbo)
- ~ 132 "SM" units
  - 128 SP "cores" each (FMA)
  - 64 DP "cores" each (FMA)
  - 4 "Tensor Cores" each
  - 2:1 SP:DP performance



- ~ 34 TFlop/s DP peak (FP64 turbo)
- 50 MiB L2 Cache
- 80 GB HBM3
- MemBW ~ 3300 GB/s (theoretical)
- MemBW ~ 3000 GB/s (measured)



# Trading single thread performance for parallelism: *GPGPUs vs. CPUs*

GPU vs. CPU light speed estimate	Control	ALU ALU	ALU ALU			
(per processor chip)	Cache					
	DRAM			DRAM		
	CPU			GPU		
	2 x AMD EPYC 9654 "Genoa"			NVidia Tesla H100 SXM "Hopper"		
Cores@Clock	2 x 96 @ 2.4 GHz			132 SMs @ ~2.0 GHz		
FP32 Performance/core	76.8 GFlop/s			~ 256 GFlop/s		
Threads@STREAM	~ 24			~ 100000		
FP32 peak	14.7 TFlop/s			~ 67 TFlop/s		
Stream BW (meas.)	2 x 360 GB/s			~ 3000 GB/s		
Transistors / TDP	~ 2x 80 (?) Billion / 2x 360 W			80 Billion/700 W		

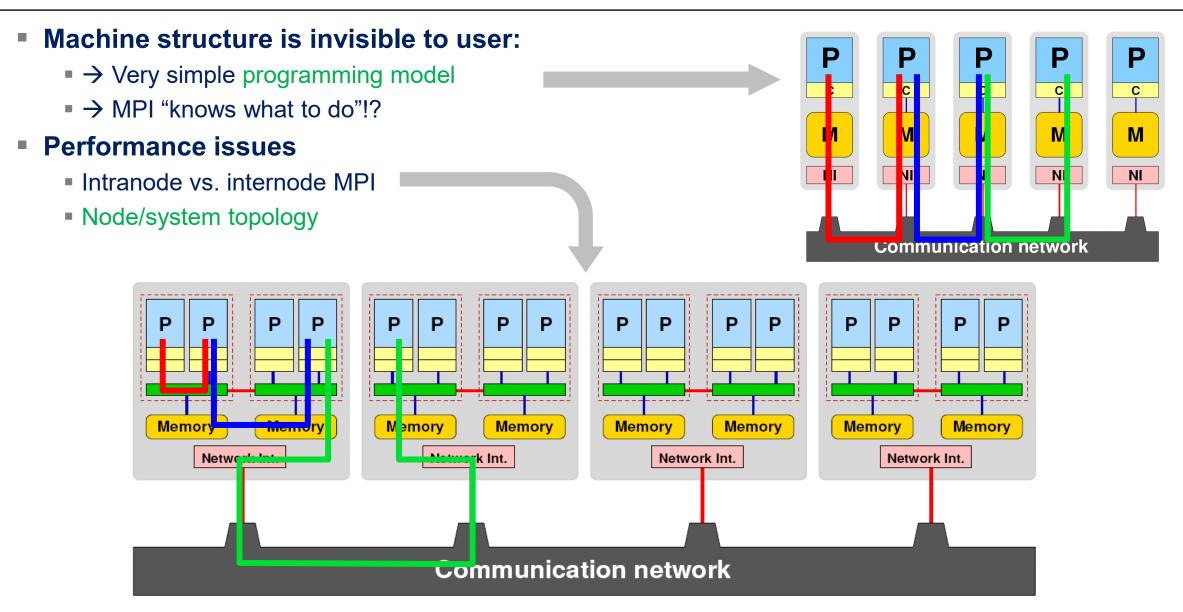




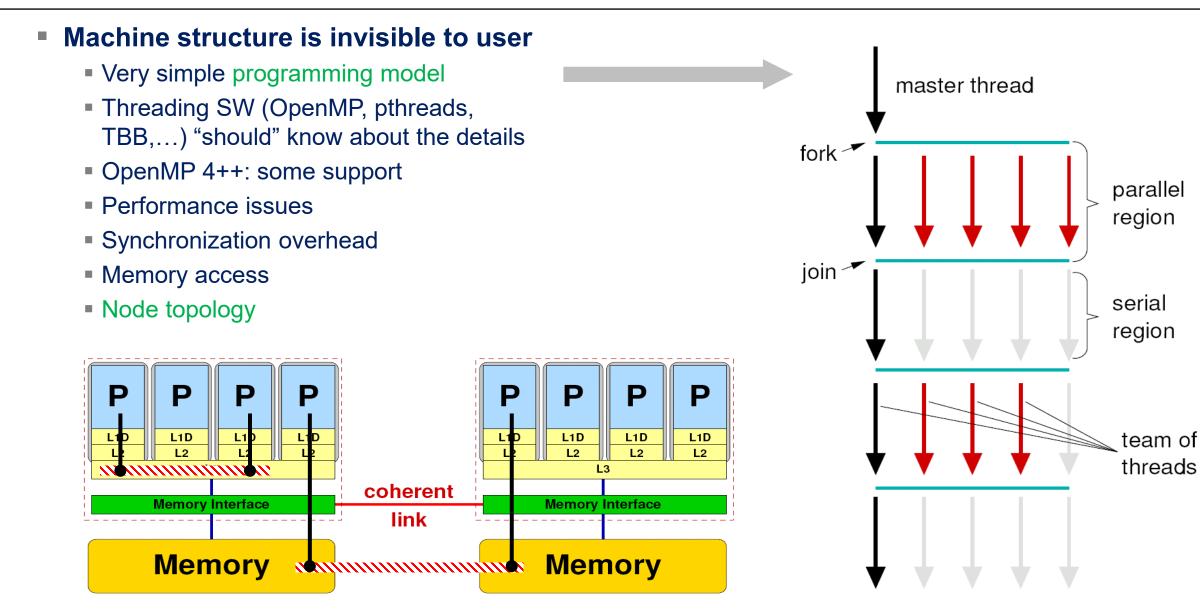
# Node topology and programming models



## Parallel programming models: Pure MPI



## Parallel programming models: Pure threading



#### **Conclusions about architecture**

- Performance is a result of
  - How many instructions you require to implement an algorithm
  - How efficiently those instructions are executed on a processor
  - Runtime contribution of the triggered data transfers
- Modern computer architecture has a rich "topology"
- Node-level hardware parallelism takes many forms
  - Sockets/devices CPU: 1-4 or more, GPGPU: 1-8
  - Cores moderate (CPU: 20-128, GPGPU: 10-100)
  - SIMD moderate (CPU: 2-16) to massive (GPGPU: 10's-100's)
  - Superscalarity (CPU: 2-6)
- Performance of programs is sensitive to architecture
  - Topology/affinity influences overheads of popular programming models
  - Standards do not contain (many) topology-aware features
    - Things are starting to improve slowly (MPI 3.0, OpenMP 4.0)
  - Apart from overheads, performance features are largely independent of the programming model