

# Building Blocks: Hardware

## Processors, Cores, Memory and Accelerators

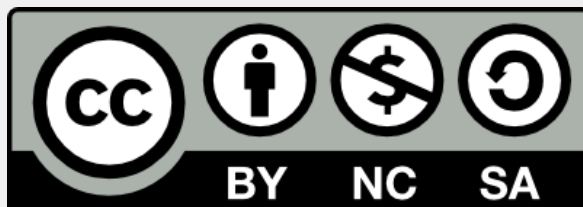
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# Outline

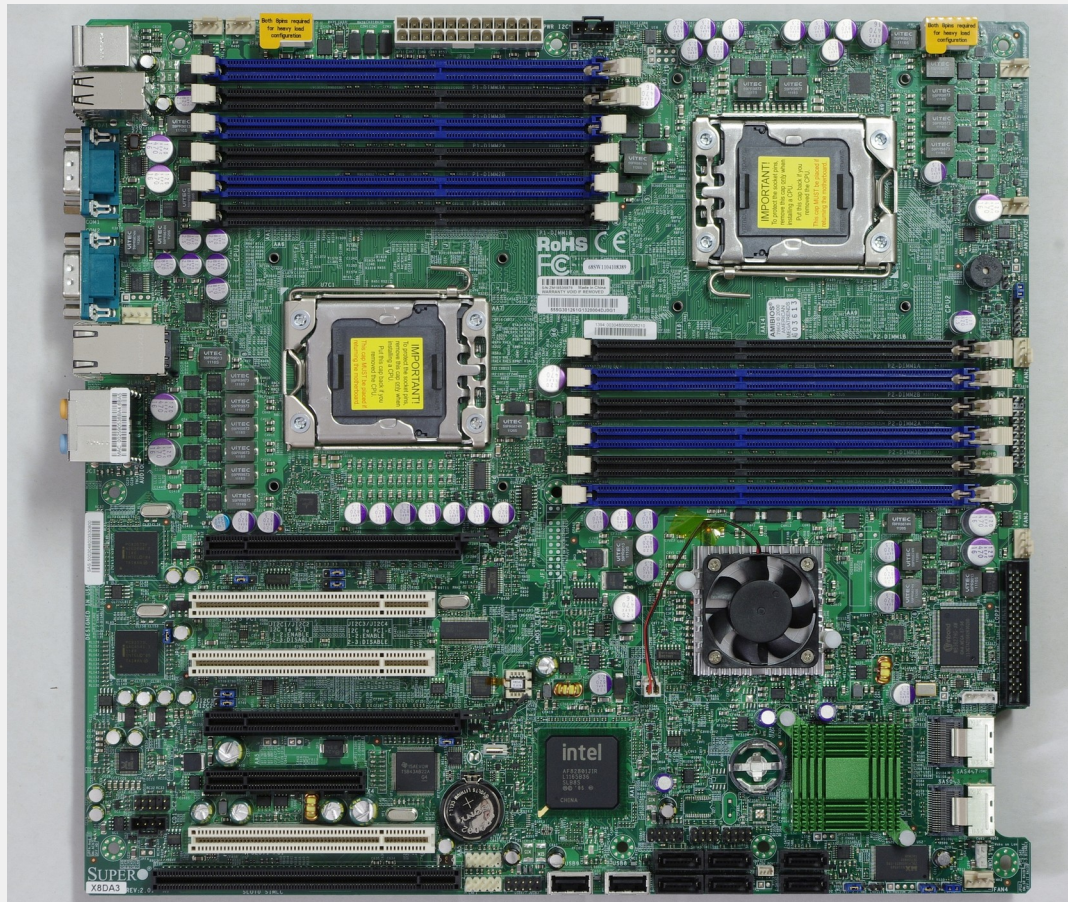
## Building Blocks of HPC systems

- Processors
  - Memory
  - Interconnect
  - Storage
- 
- Evolution of the Processor
    - Moore's Law
    - Parallelism in Hardware
      - Vector instructions (SIMD)
      - Multicore processors
      - Simultaneous multi-threading (SMT)
  - Accelerators (GPU)
    - What are they good for?

# What is a computer?



# What is a computer?

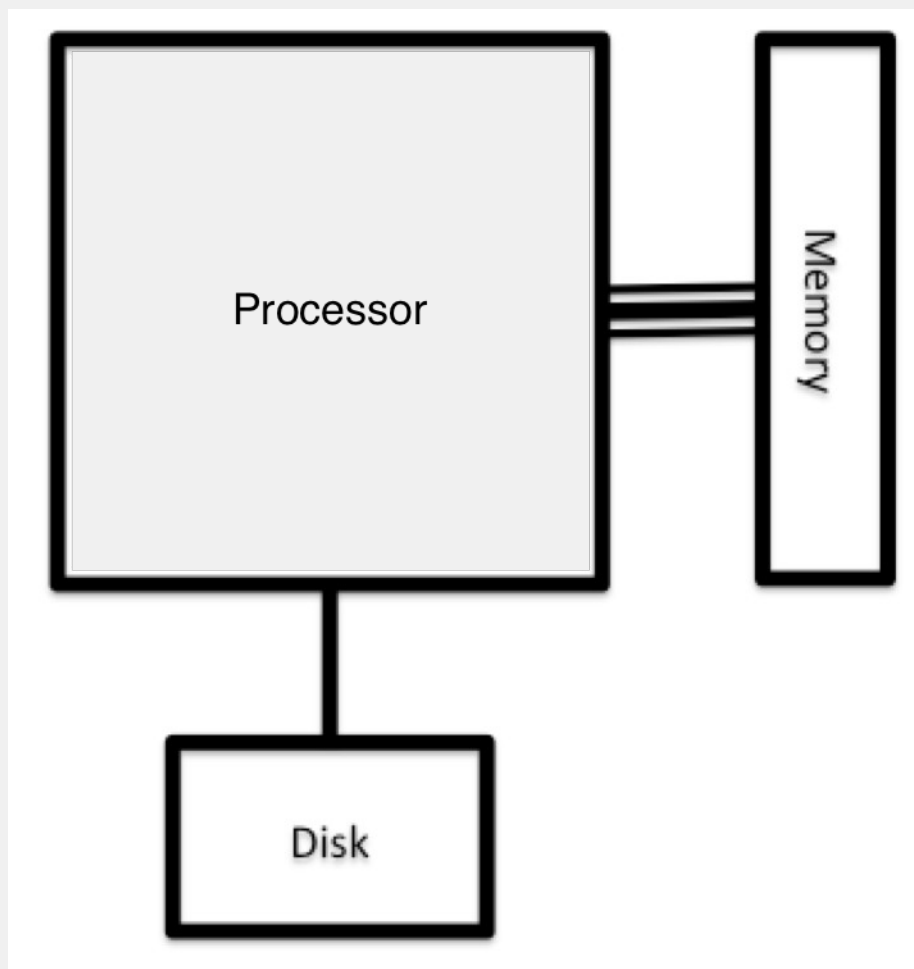


# Computational Building Blocks

Four principal hardware technologies make up HPC systems:

- Processors (& accelerators)
  - to calculate
- Memory
  - for temporary storage of data
- Interconnect
  - enabling processors to talk to each other (and the outside world)
- Storage
  - disks for storing input/output data, other for long term archiving of data
- We will focus on the first two of these

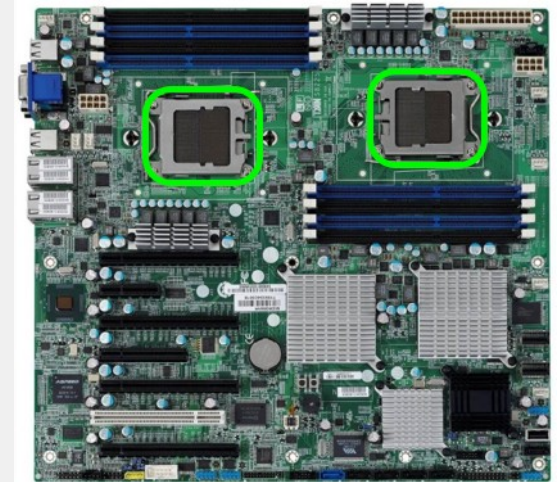
# Anatomy of a Computer





## What do you mean, “processor”?

- Terminology gradually varied over time and in different contexts (hardware, software)
- Usually taken to mean “the thing you plug in to a socket on the motherboard” (e.g. two processor sockets below)
- MareNostrum: each node has two processors (each in its own socket)
- Your laptop: one multicore processor



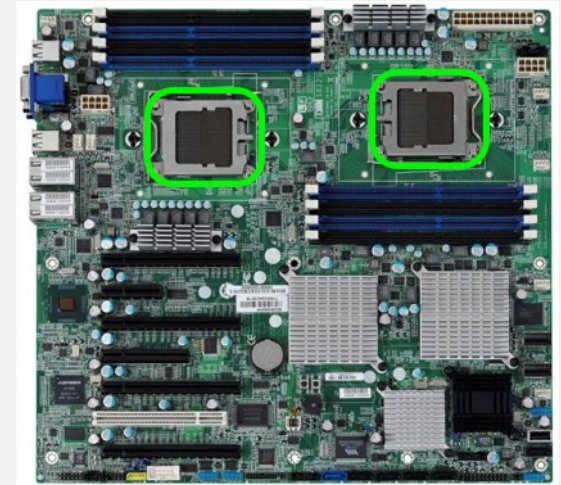


# What do you mean, “CPU”?

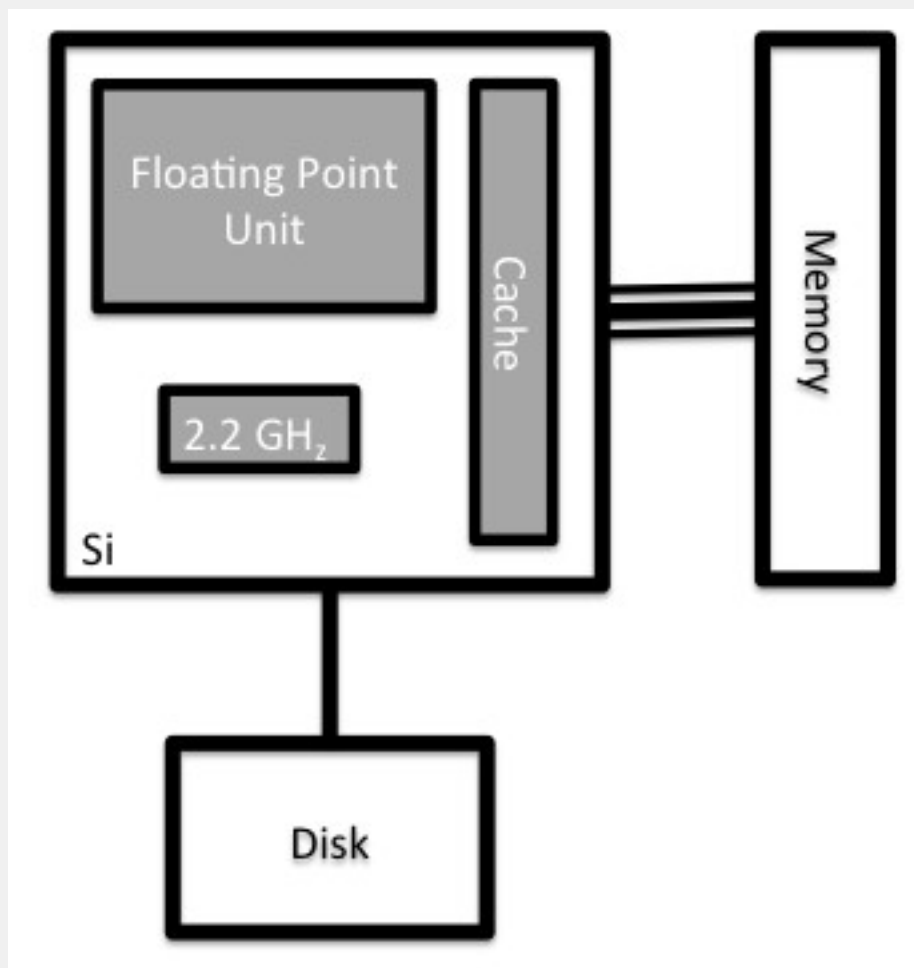
Ambiguous!

Nowadays may refer to:

- Entire multicore processor
  - Usage from single-core processor era
- A (physical) processor core
- A logical core
  - Anything that looks to the Operating System like an independently addressable and usable core to assign threads or processes to execute on
  - E.g. physical cores, also logical subunit(s) of physical cores



## Anatomy of a Computer (single-core processor)



# Processors

- Basic functionality
  - execute instructions to perform arithmetic operations (integer and floating point)
  - load data from memory and store data to memory
  - decide which instructions to execute next
- Mathematical operations performed on values in registers (local storage in the processor)
  - Moving data between memory and registers = load and store instructions
  - Separate integer and floating point registers
  - Typical size ~100 values
- Basic characteristics:
  - Clock speed
  - Peak floating point capability

# Functional Units

Functional units are the basic building blocks of processors

- Number and type of units varies with processor design.

Basic units for any processor include:

- **Instruction unit**

- Responsible for fetching, decoding and dispatching of instructions.
- Fetches instruction from instruction caches
- Decodes instruction.
- Sends the instructions to the appropriate unit
- May also be responsible for scheduling instructions (see later).

# Functional Units

Basic units for any processor include:

- **Integer unit**

- Handles integer arithmetic
- Integer addition, multiplication and division
- Logical ops (and, or, shift etc.)
- Also known as arithmetic and logic unit (ALU)

- **Floating point unit**

- Handles floating point arithmetic
- Addition, multiplication, division.
- Usually the critical resource for HPC
- Machines sold by peak floating point operations per second (flop/s)

# Functional Units

Basic units for any processor include:

- **Control unit**

- Responsible for branches and jumps

- **Load/store unit**

- Responsible for loading data from memory and storing it back.

- **Register file**

- Local storage in the CPU
  - Accessed by name (not address)
  - Separate files for integers/addresses and floating point

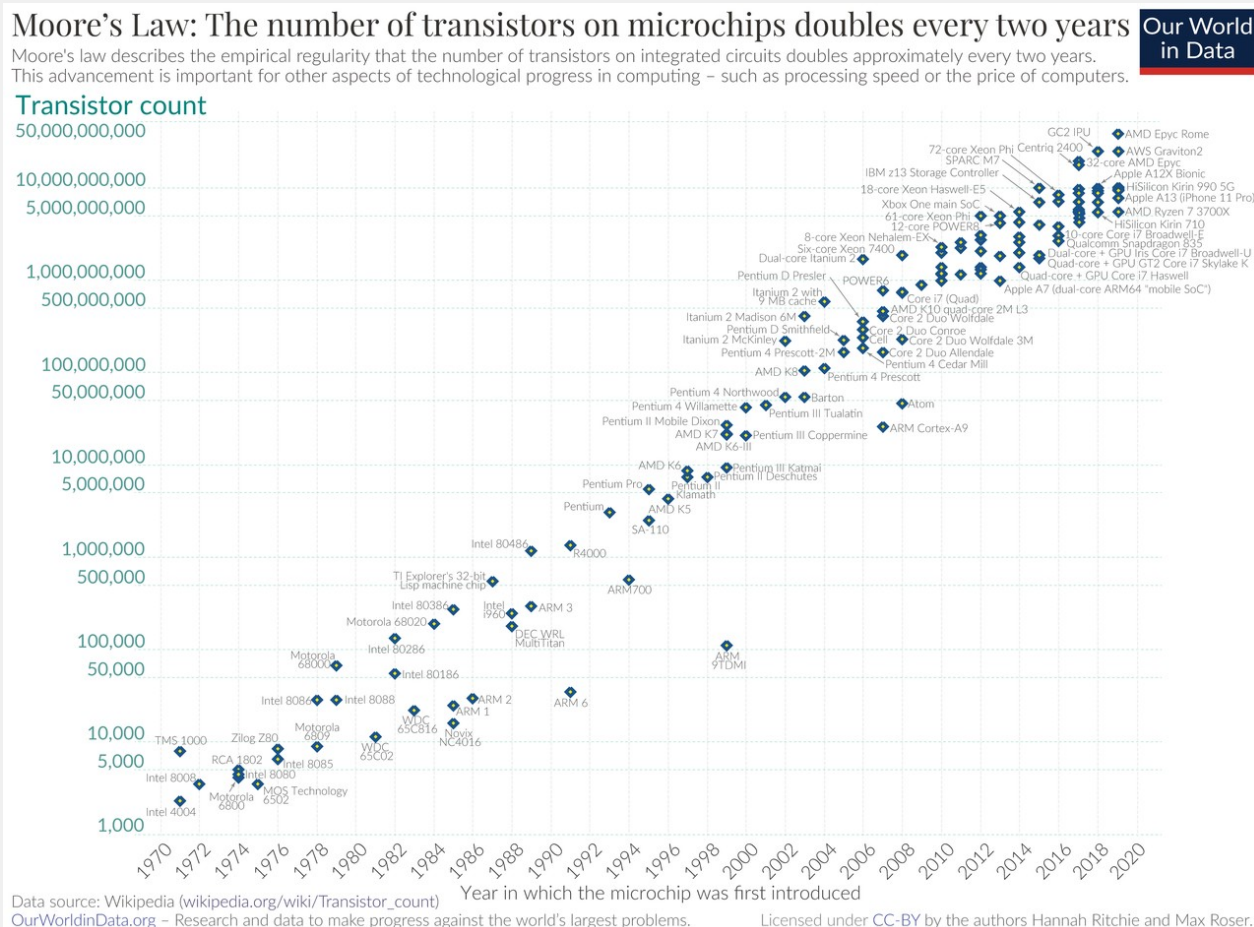
Also memory management, cache controller, bus interface, graphics/multimedia,.....

# Moore's Law, Processor Evolution, and Parallelism in Hardware



# Moore's Law

- Number of transistors doubles every 18-24 months
  - enabled by advances in semiconductor technology and manufacturing processes



<https://ourworldindata.org/uploads/2020/11/Transistor-Count-over-time.png>

# What to do with all those transistors?

- For over 3 decades (the “good old days”) until early 2000’s
  - processors became more complicated / sophisticated
  - caches became bigger
  - clock speeds increased year on year (100 MHz, 200 MHz, 400MHz, ...)
  - for your program to run faster just wait a year and buy a newer processor
- Clock rate increases as inter-transistor distances decrease
  - so performance doubled every 18-24 months
- Came to a grinding halt about a decade ago
  - reached power and heat limitations
  - who wants a laptop that runs for an hour and scorches your trousers!

# Processor performance

- Clock speed determines rate at which instructions are executed
  - modern chips are around 2-3 GHz
  - integer and floating point calculations can be done in parallel
  - can also have multiple issue, e.g. simultaneous add and multiply
  - peak flop rate is just clock rate x no. of floating point operations per clock cycle
  
- Decades of research and \$\$\$ into hardware innovations, complex design and optimisations
  - Out-of-order execution
    - Assembly code specifies an order of instructions....
    - Hardware chooses to reorder instructions to minimise pipeline stalls
    - Requires some complex bookkeeping to ensure correctness
  - Branch prediction
    - Hardware tries to guess which way the next branch will go
    - Uses a hardware table that tracks the outcomes of recent branches in the code.
    - Keeps the pipeline going and only stalls if prediction is wrong

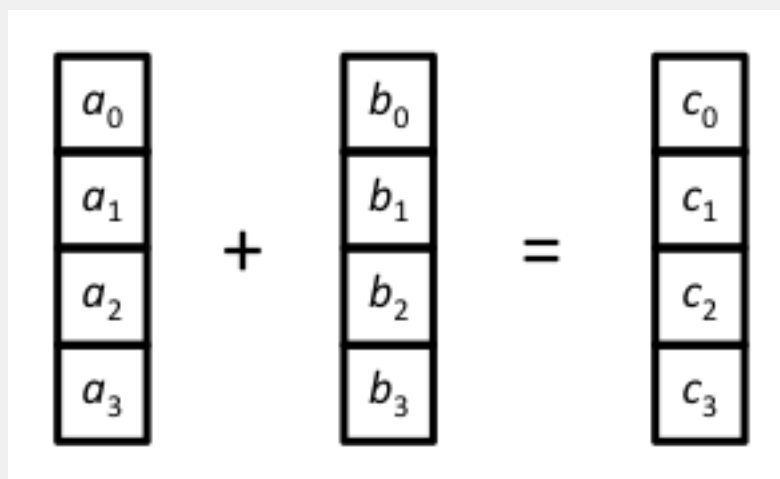
# Processor performance

Introduce parallelism into the processor itself

- **Vector** instructions (**vectorisation**) (“**SIMD**” = “Single Instruction, Multiple Data”)
- Multiple cores close together on same chip (**multicore processor**)
- Simultaneous Multi-Threading (“**SMT**”)

# Single Instruction Multiple Data (SIMD)

- For example, vector addition:

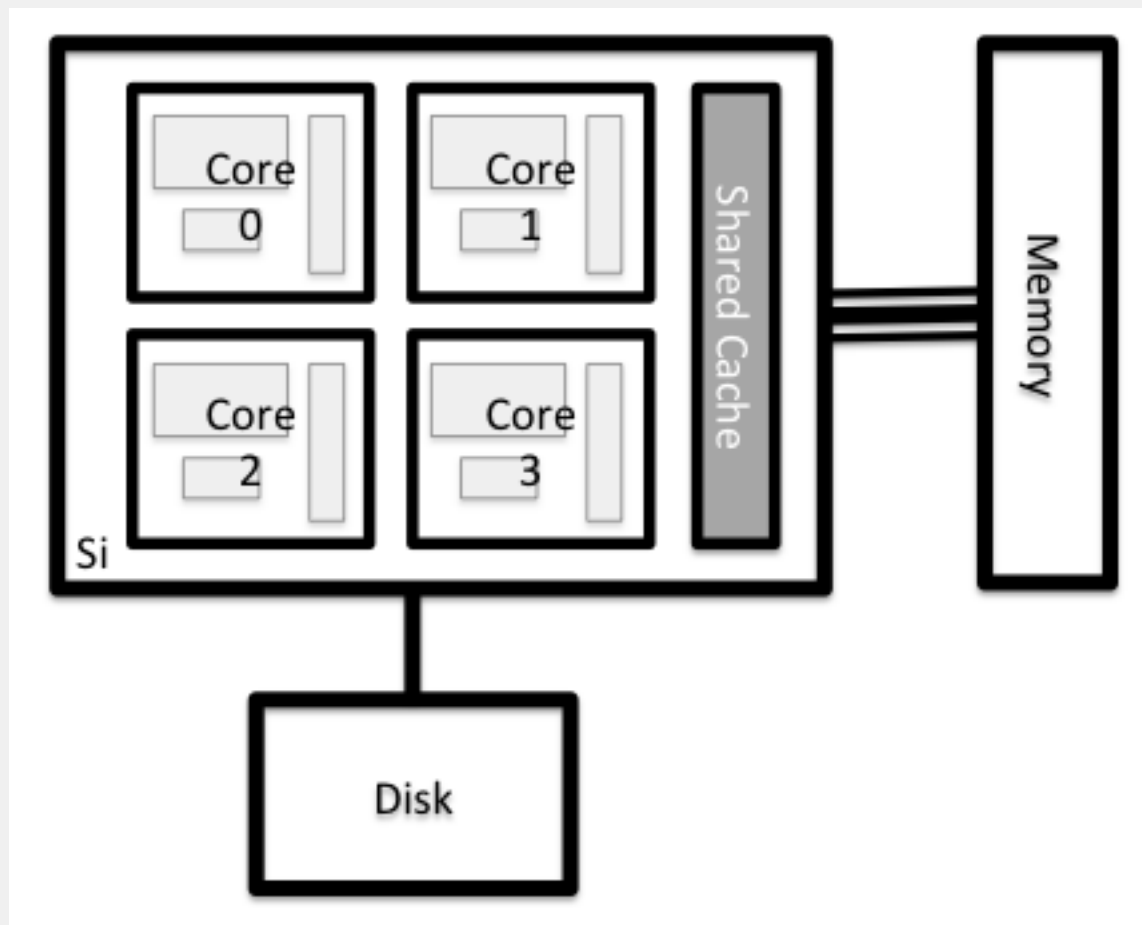


- single instruction adds 4 numbers
- potential for 4 times the performance
- SIMD width (vector width) is the number of operations encoded in a SIMD instruction
  - Often 2, 4 or 8

# Multicore

- Twice the number of transistors gives 2 choices
  - a new more complicated processor with twice the clock speed
  - two versions of the old processor with the same clock speed
- Second option is more power efficient
  - and now the only option as we have reached heat/power limits
- Effectively two independent single-core processors
  - ... except they can share cache
  - Commonly called “cores”
  - Cores may also also share some functional units

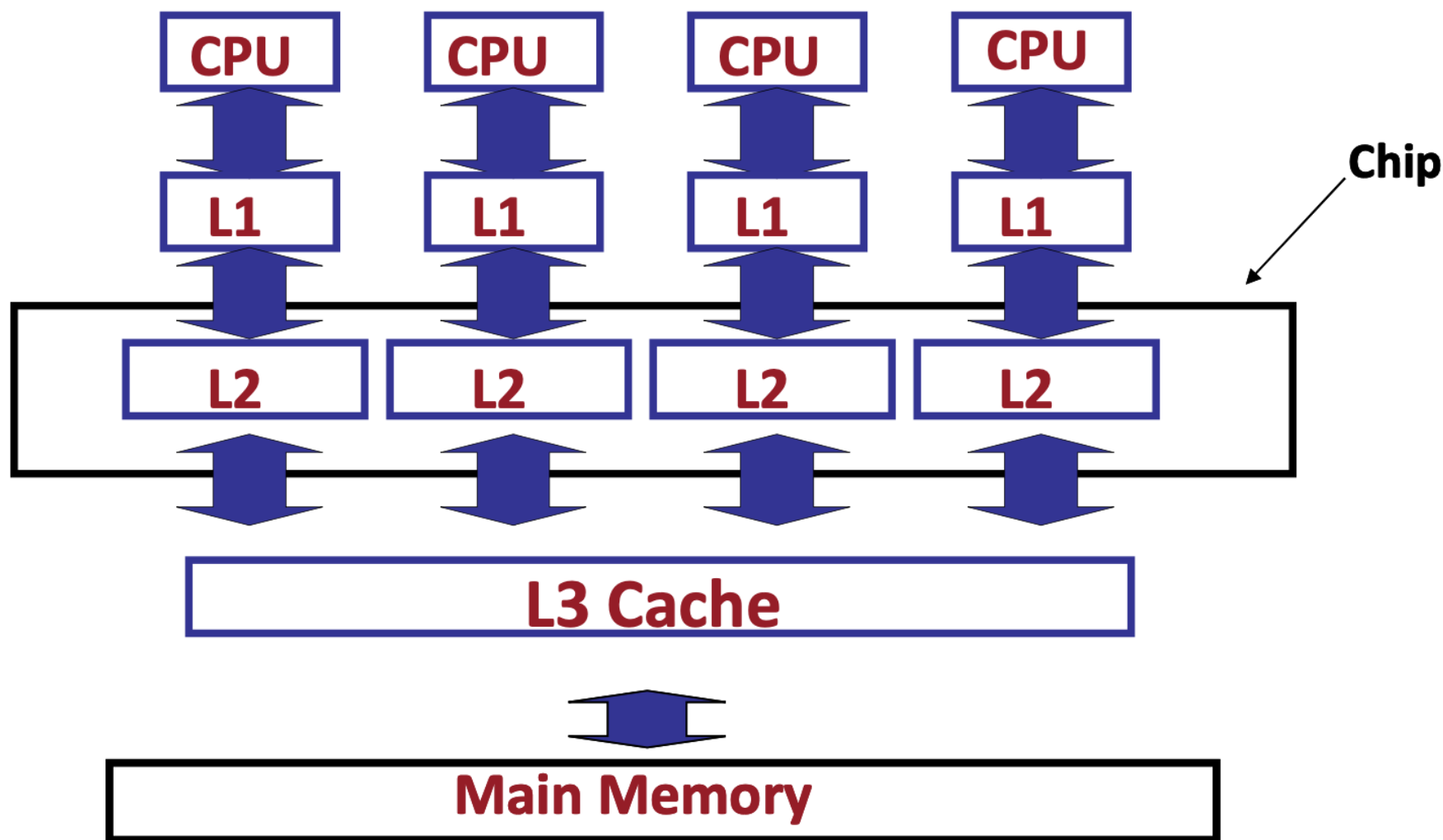
# Anatomy of a computer (single processor, multicore)



- Cores share path to memory
  - More cores makes this an *increasing* bottleneck!



# Multicore Cache Hierarchy



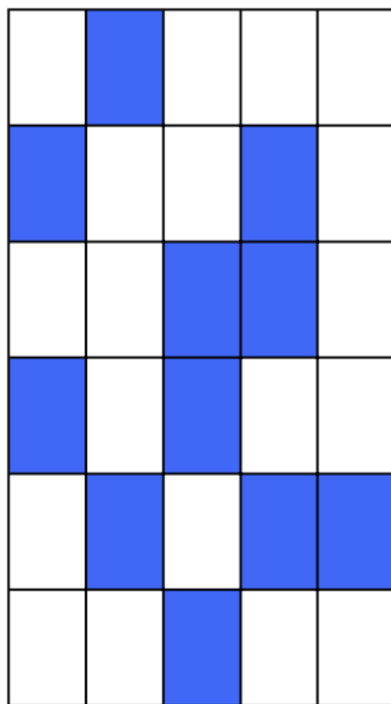
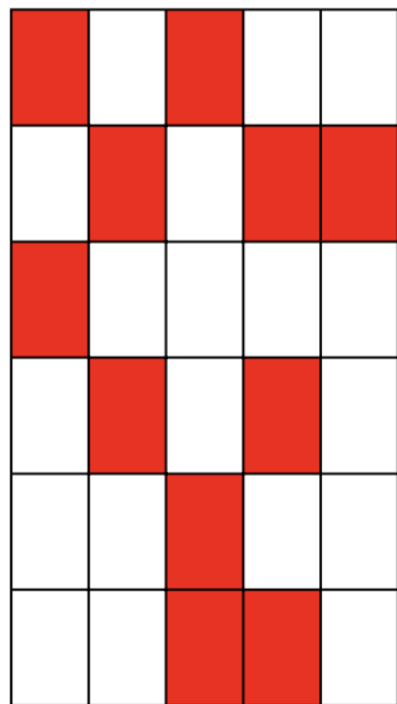
# Multicore Cache Hierarchy

- Cores on the same chip can communicate with low latency and high bandwidth
  - reads and writes through shared cache
- Cores share space in the shared cache
  - Possible contention: one thread may suffer capacity and/or conflict issues caused by threads/processes on another core
  - If only single core is running, then it may have access to the whole shared cache
- Cores also share off-chip bandwidth
  - access to main memory

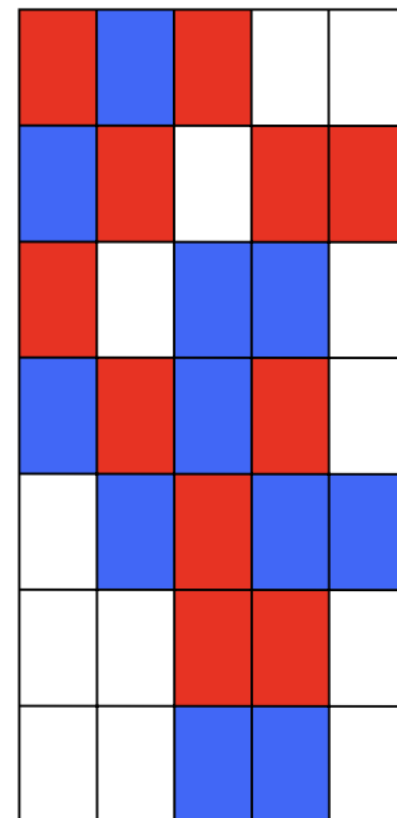
# Simultaneous Multi-Threading (SMT)

- Some processors supports running multiple instruction streams simultaneously on the same processor, e.g.
  - stream 1: loading data from memory
  - stream 2: multiplying two floating-point numbers together
- Requires some replication of hardware, but everything else is shared between threads
  - functional units, register file, memory system (including caches)
- “Threading” can be a misnomer - can refer to processes as well as threads
  - These are “hardware threads”, not software threads
    - = ability to execute instructions threads or processes
  - Appear to the Operating System as distinct logical cores
  - For most architectures, two or four threads is all that makes sense
    - Intel Xeon & AMD EPYC Zen2 supports 2-way SMT

# Simultaneous Multi-Threading (SMT)



**Time**



Two threads on two physical cores

Two threads on one physical core  
with SMT enabled

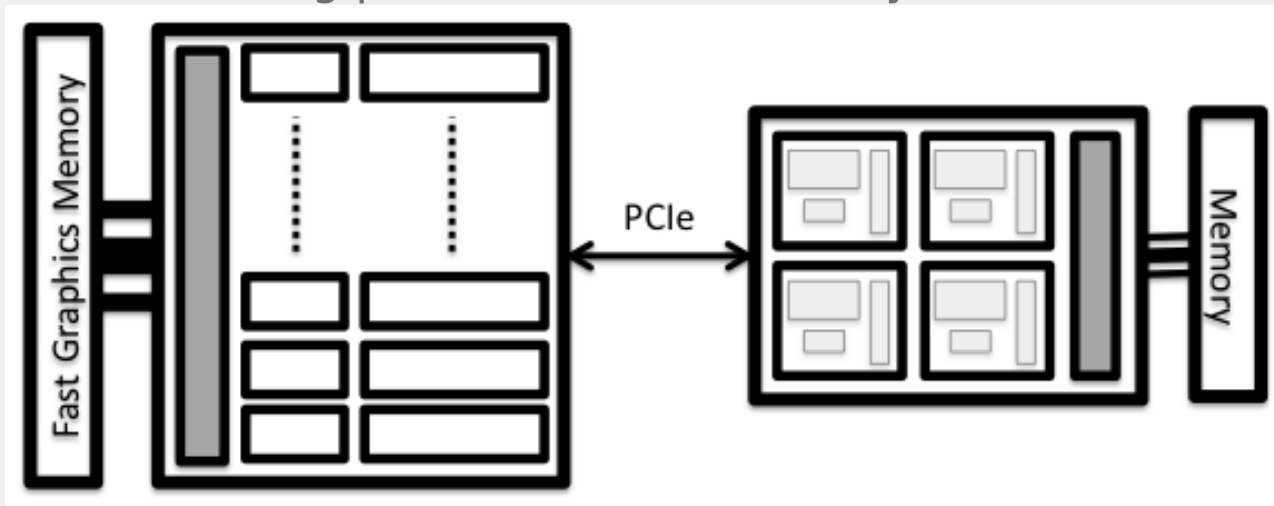
# Simultaneous Multi-Threading (SMT)

- How successful is SMT?
  - depends on the application, and how the threads contend for the shared resources.
- In practice, gains seem to be limited to around 1.2 to 1.3 times speedup over a single thread
  - benefits will be limited if both threads are using the same functional units (e.g. FPU) intensively.
- For some codes, SMT can cause slow down
  - increased contention for memory bandwidth and/or cache space
  - increasing the number of threads/processes can increase overheads such as communication or load imbalance

# Accelerators

# Anatomy

- An *Accelerator* is an additional resource that can be used to off-load heavy floating-point calculation
  - additional processing engine attached to the standard processor
  - has its own floating point units and memory



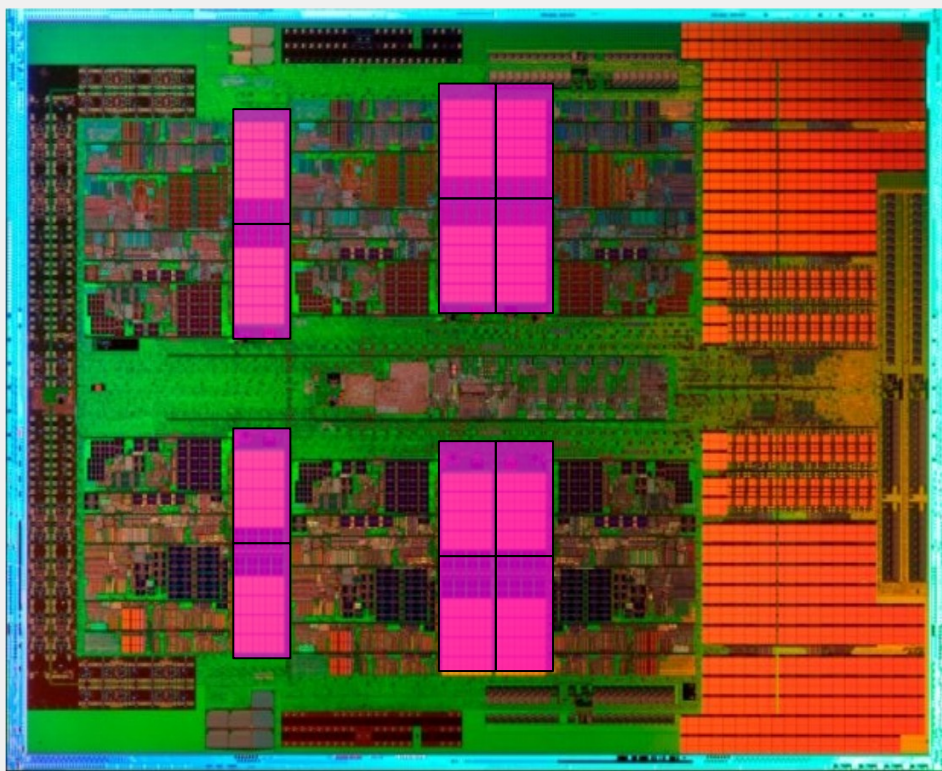


# Accelerators

- Very widespread to include on compute nodes alongside the main CPU
  - large numbers of relatively simple cores
  - high memory bandwidth
  - separate memory from CPU
  
- Much of current interest is focussed on GPGPUs (general purpose graphics processing units)
  - low cost due to high mass market volumes
  - tricky to program
  - significant overheads in moving data to/from GPU memory

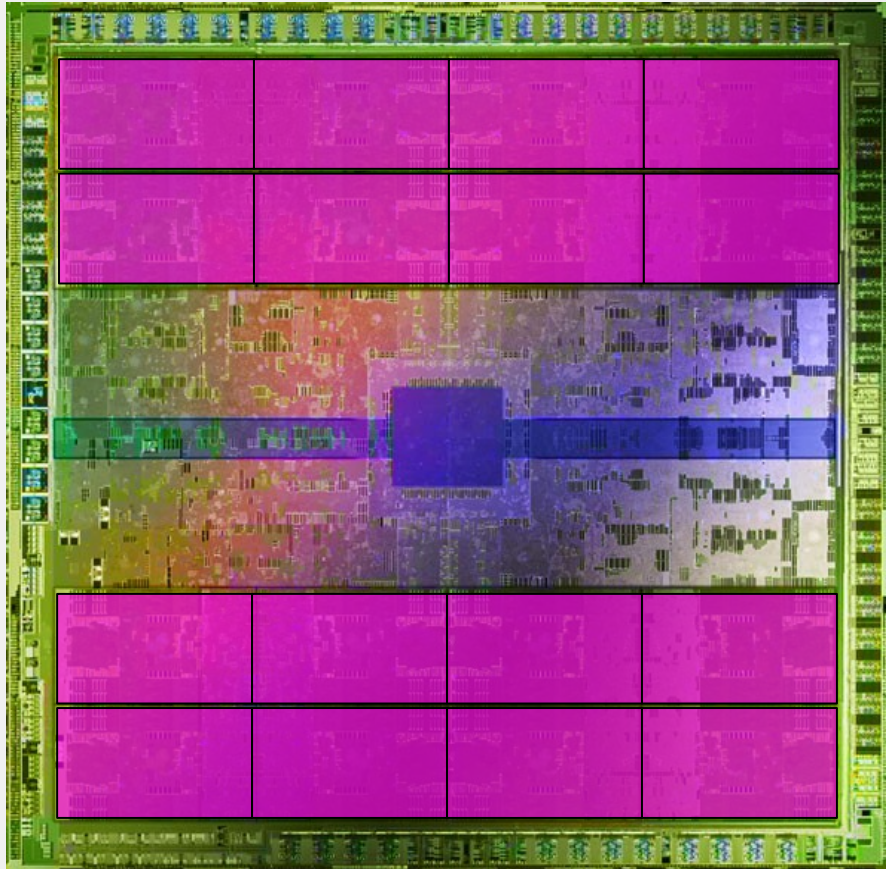
# AMD 12-core CPU

- Not much space on CPU is dedicated to computation



= compute unit  
(= core)

# NVIDIA Fermi GPU



- GPU dedicates much more space to computation
  - At expense of caches, controllers, sophistication etc



= compute unit  
 (= SM  
 = 32 CUDA cores)

# Memory

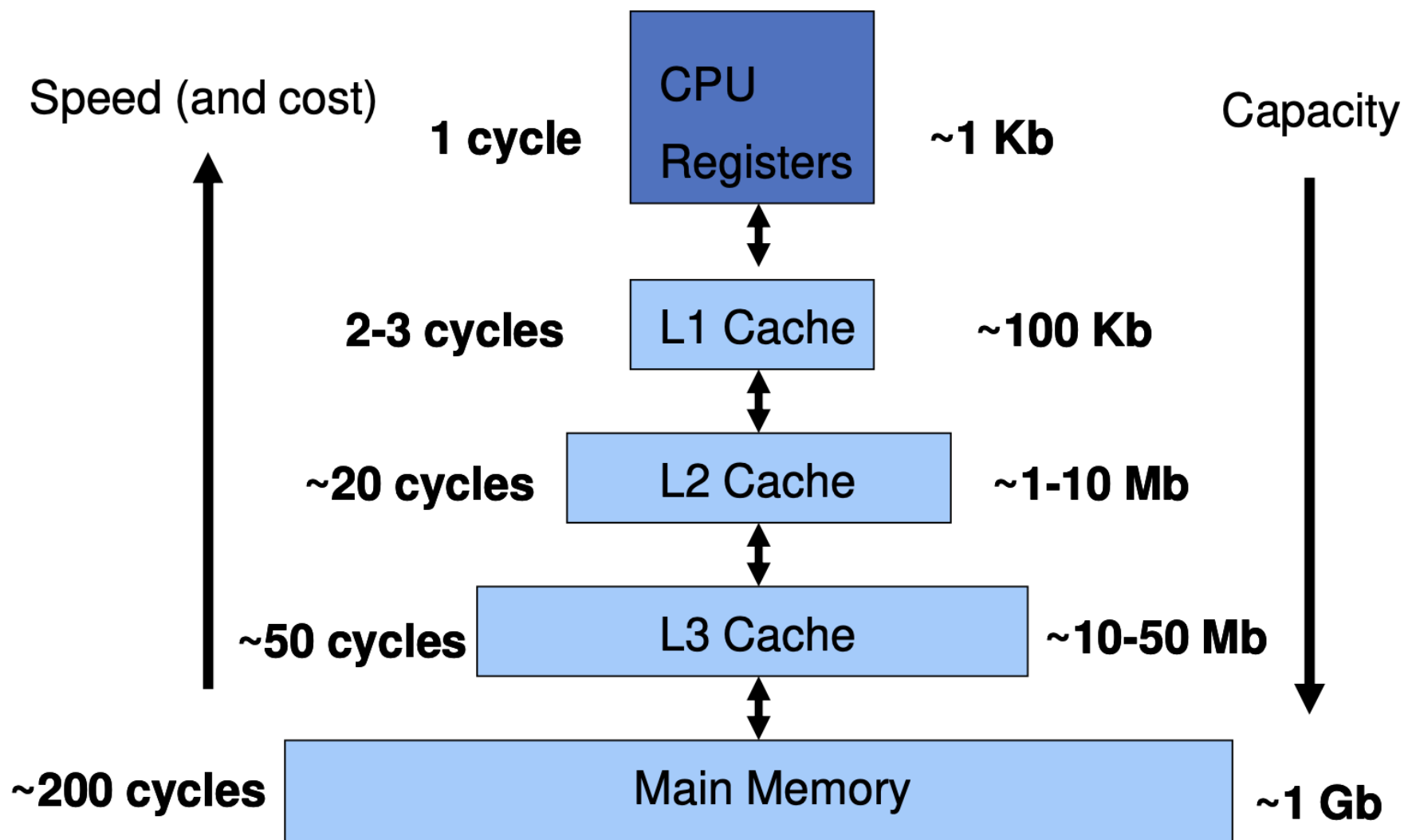


# Data Access Bottlenecks

Performance depends on getting data to the processor quickly

- Latency: time delay until data starts arriving
- Bandwidth: amount of data arriving per second
  
- Disk access is slow
  - few hundred MB/s
  - significantly higher latency than memory
  
- Memory access is faster than disk
  - Large enough memory may contain all application data
  - Can still be too slow - few tens of GB/s
  
- Accessing cache (fast memory inside processor) is much faster
  - hundreds of GB/s
  - limited in size: a few MB at most

# Memory Hierarchy



# Performance (overview)

Application performance often described as:

- Compute bound (limited by processor performance)
- Memory bound (limited by memory access)
- IO bound (limited by disk access)
- Communication bound (limited by interconnect)

For current HPC codes:

- many calculations are limited by memory bandwidth
- processor can calculate much faster than it can access data