

Introduction to High-Performance Computing

Session 07 Performance Optimization



Performance Modelling

• the following slides are based on

https://moodle.rrze.uni-erlangen.de/course/view.php?id=311

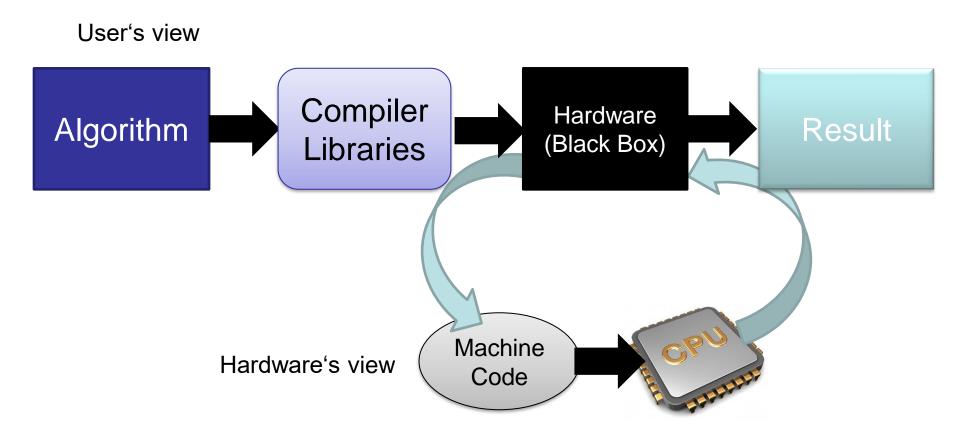
- 2-day course during MCS Summer School 2014 given by Georg Hager
- Book: G. Hager and G. Wellein:

Introduction to High Performance Computing for Scientists and Engineers,

CRC Computational Science Series, 2010. ISBN 978-1439811924 http://www.hpc.rrze.uni-erlangen.de/HPC4SE/



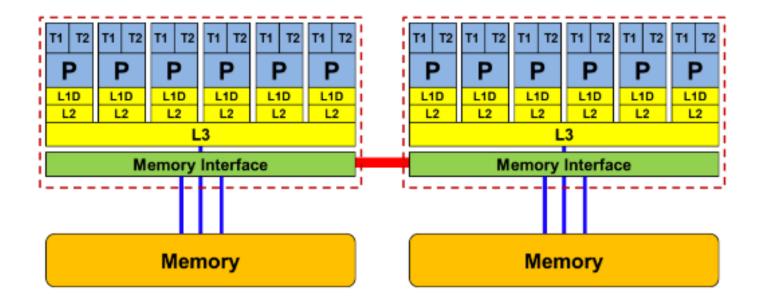
Computer Software and Hardware



16.03.2021



Modern Computer Architecture

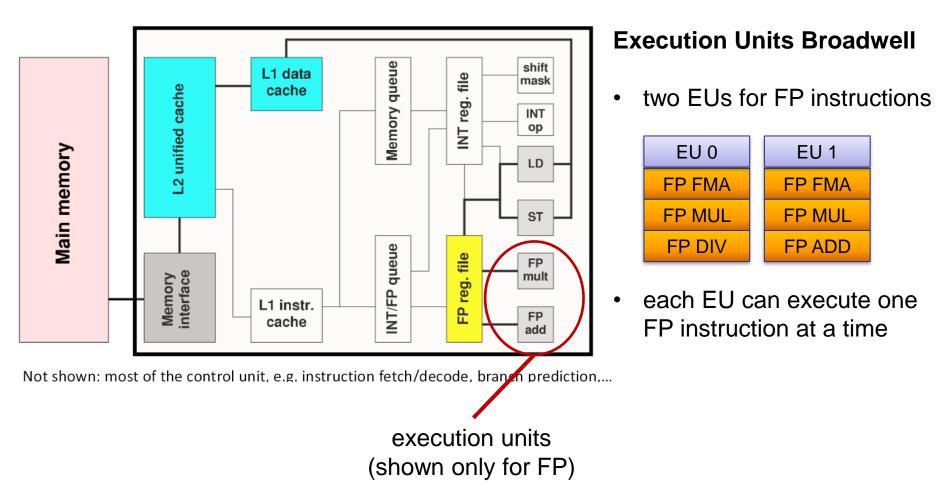


- today: dual-socket node
 - multiple cores per socket/CPU
 - ccNUMA architecture
 - socket interconnect





Detailed View Compute Core





Example: Divide Throughput

• in the **Pi.cpp** code the function **f(x)** has one division

```
// define f so that integral of f from 0 to 1 is Pi
double f(const double x) {
   return (4.0/(1.0+x*x));
}
```

- division is the dominant operation (other instructions can be hidden)
- for **n** evaluations of **f** we get $t = n \cdot \frac{c}{v}$
- Broadwell CPUs need c = 5 cycles/division (throughput) and assuming turbo mode (clock speed $\nu = 2.5$ GHz) we would expect t = 0.2s for $n = 10^8$



Execution of Instructions

 programmer's view: 	 hardware's view:
for (int i=0; i <n; i++)<br="">A[i] = A[i] + B[i];</n;>	load r1 = A(i) load r2 = B(i) add r1 = r1 + r2
 user work: N Flops (ADDs) 	<pre>store A(i) = r1 inc i branch top if i<n< pre=""></n<></pre>

programm performs computation, FLOP is the basic work done processor executes instructions, instructions is the basic work done

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Basic Compute Resources

- instruction execution
 - primary resource for computations, hardware is designed to increase instruction throughput as much as possible
 - difficult for general purpose computing, what is a typical workload?
- data movement
 - consequence of instruction execution
 - in the example two loads and one store (double 24 byte)

What is the bottleneck of an application?



Flop/s vs. Memory Bandwidth

- a floating-point operation (Flop) is the basic unit of work
 - theoretical peak performance Intel Xeon E5-2650 v4

 $P_{\text{peak}} = 422.5 \text{ GFlop/s}$

- equivalent to $16 \operatorname{Flop}/(\operatorname{core} \cdot \operatorname{cy})$
- memory bandwidth
 - maximum for Intel Xeon E5-2650 v4 is 76.8 GB/s

(https://ark.intel.com/products/91767/Intel-Xeon-Processor-E5-2650-v4-30M-Cache-2_20-GHz)

equivalent to

(more info: http://sites.utexas.edu/jdm4372/tag/memory-bandwidth/)

³⁵ Byte/cy



Example Bandwidth Limited Execution

consider the vector-triad

for (j=0; j<STREAM_ARRAY_SIZE; j++) a[j] = b[j]+scalar*c[j];</pre>

- included in the STREAM benchmark (see https://www.cs.virginia.edu/stream/)
- 2 Flop/iteration and 24 Byte/iteration
- at 16 Flop/cy on a single core 192 Byte/cy are needed

➔ memory bandwidth is the limiting factor here



STREAM Benchmark

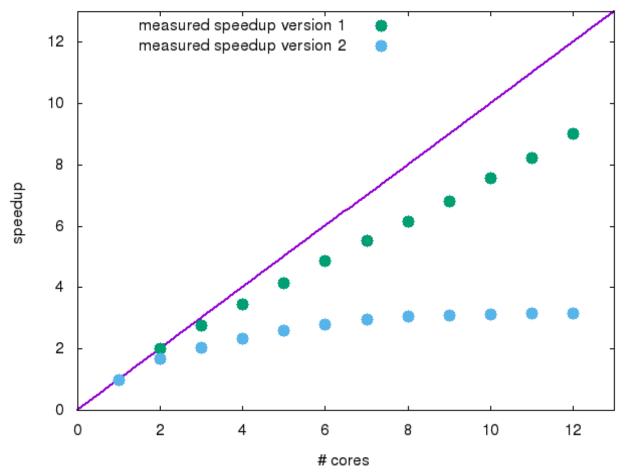
https://www.cs.virginia.edu/stream/

- simple tool to measure memory bandwidth
 - timing of bandwidth-limited vector operations
- some results
 - single core bandwidth is about 20 GB/s
 - maximum bandwidth measured is about 64 GB/s per socket and 128 GB/s per node (two sockets)
 - about half of the cores are need to get (close to) maximum bandwidth



Parallel Speedup

version 1, very good scaling version 2, almost no scaling







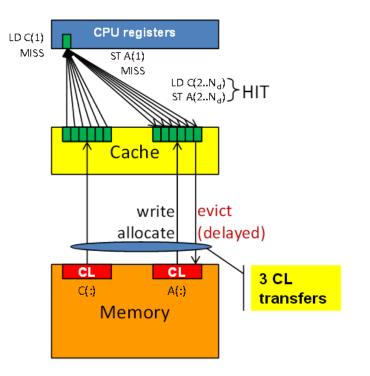
• 3d "Stencil" update (Jacobi)

note that the order of the loops is important (and depends on the ordering of multi-dimensional arrays in memory)



Memory Access Patterns

- 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)
- MISS: Load or store instruction does not find the data in a cache level
 CL transfer required
- Example: Array copy A(:)=C(:)





Hardware Locality

- compute nodes are increasingly complex
 - ccNUMA architectures
- the hwloc library provides some tools to

(https://www.open-mpi.org/projects/hwloc/)

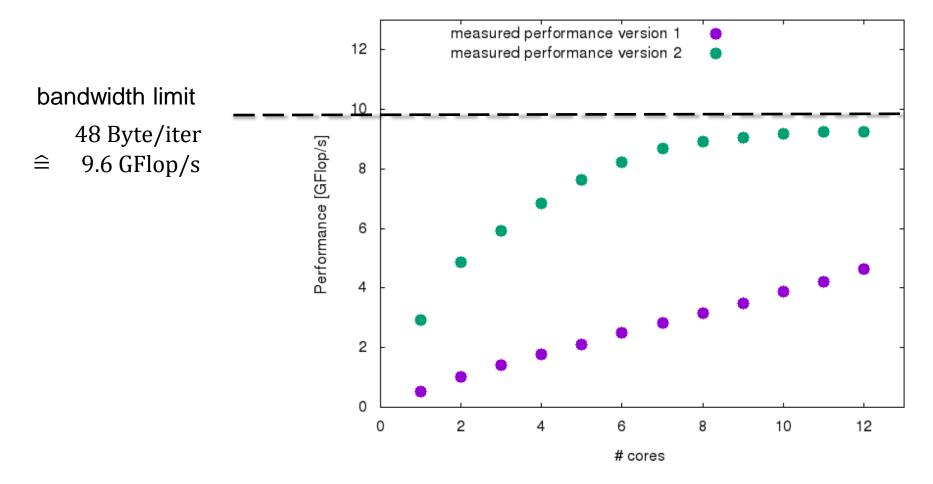
- obtain information about the node topology (lstopo)
- bind processes to specific cores/sockets/...
- binding/pinning of threads may improve performance (hwloc-bind ... <command>)
- difficult to decide, e.g. is it better to use neighboring cores or different sockets?

Package P#0											
L3 (30MB)											
L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KE
L1d (32KB)	L1d (32KB)	Lld(32KB)	L1d (32KB)	Lld(32KB)	Lld(32KB)	L1d (32KB)	Lld(32KB)	L1d (32KB)	L1d(32KB)	L1d (32KB)	L1d (32KE
L1i (32KB)	L11(32KB)	L11(32KB)	L11 (32KB)	L11(32KB)	L11(32KB)	L11(32KB)	L11(32KB)	L11(32KB)	L11(32KB)	L11(32KB)	L11(32KB
Core P#0	Core P#1	Core P#2	Core P#3	Core P#4	Core P#5	Core P#S	Core P#9	Core P#10	Core P#11	Core P#12	Core P#13
PU PW0	PU PW1	PU PW2	PU PV3	PU 9#4	PU P#5	PU 996	PU PW7	PU PW8	PU P#9	PU P#10	PU P#11
PU PW0	PU Pv1	PU PW2	PU PV3	PU P#4	PU P#5	PU P#6	PU PW7	PU P#8	PU P#9	PU P#10	PU P#11
PU P#0		PU PW2	PU PV3	PU P#4	PU PW5	PU P#6	PU PW7	PU Pws	PU P#9	PU Pw10	PU P#11
		PU PW2	PU PV3	PU PM	PU PW5	PU P#6	PU PW7	PU PV3	PU Pit9	PU P#10	PU PU11
NUMANode P#		PU PW2	PU PV3	PU PM4	PU P#5	PU PP6	PU PW7	PU Pvs	PU Pii	PU Pe10	PU P#11
NUMANode P# Package P#1		PU PW2	PU 993	PU PH4	PU P#5	PU P96	PU PW7	PU Pv8	PU PH9	PU P#10	PU P911
NUMANode P# Package P#1 L3 (30MB)	1 (123GB)										
NUMANode P# Package P#1 L3 (30MB) L2 (256KB)	1 (123GB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	L2 (256KB)	1.2 (256KB)	L2 (256KB)	L2 (255KB)	L2 (256KB)	L2 (256KB)	L2 (256KE

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performance of version 2 is better by factor of few

Parallel Performance





How is the Hardware optimized for performance?

- speedup memory access with cache (see before)
- pipelining of arithmethic units
- instruction pipeline
- instruction level parallelism
- simultaneous multi-threading (SMT)
- SIMD processing



Pipelining

Idea:

- Split complex instruction into several simple / fast steps (stages)
- Each step takes the same amount of time, e.g., a single cycle
- Execute different steps on different instructions at the same time (in parallel)

Allows for shorter cycle times (simpler logic circuits), e.g.:

- floating point multiplication takes 5 cycles, but
- processor can work on 5 different multiplications simultaneously
- one result at each cycle after the pipeline is full

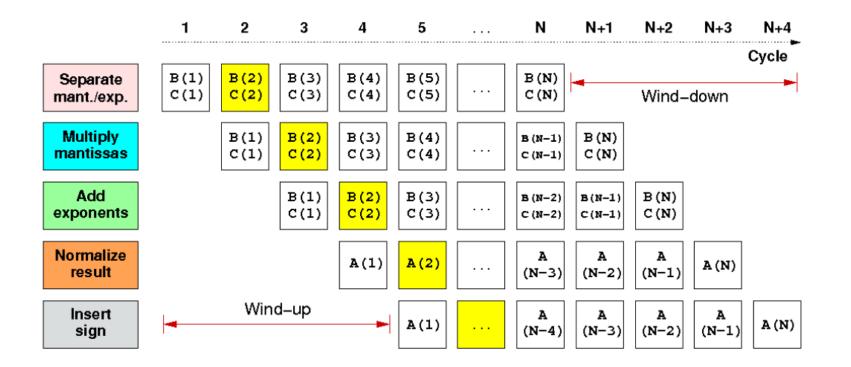
Drawback:

- Pipeline must be filled startup times (#Instructions >> pipeline steps)
- Efficient use of pipelines requires large number of independent instructions → instruction level parallelism
- Requires complex instruction scheduling by compiler/hardware softwarepipelining / out-of-order

Pipelining is widely used in modern computer architectures



Pipelinig – 5 stage Multiplication



First result is available after 5 cycles (=latency of pipeline)! Wind-up/-down phases: Empty pipeline stages

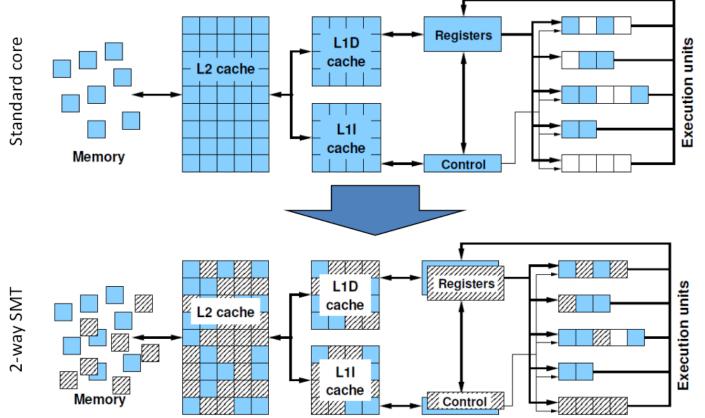
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simultaneous multi-threading (SMT)

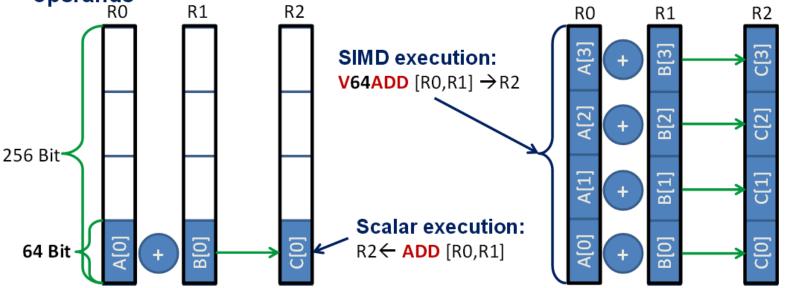
SMT principle (2-way example):





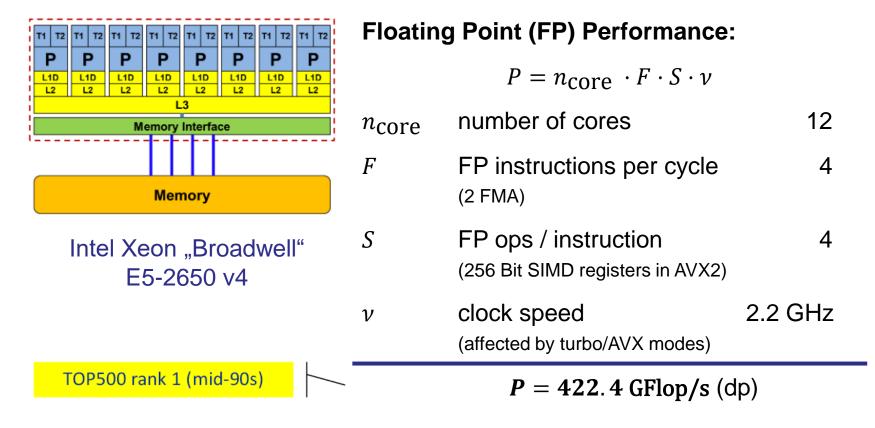
SIMD processing

- Single Instruction Multiple Data (SIMD) operations allow the concurrent execution of the same operation on "wide" registers
- 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
- Adding two registers holding double precision floating point operands R0 R1 R2 R0 R1





Processor Peak Performance



But: P = 8.8 GFlop/s for serial, non-SIMD code



Performance Bottleneck

- many floating point computation on little data
 → bound by the processing speed of the CPU
 - possibly increase number of cores
 - make use of SIMD processing
 - note: recent CPU may have lower clock speed for AVX
- few floating point operation per data
 → bound by memory bandwidth
 - change algorithm/parallelization to make better use of cache
 - increase compute intensity





- OMP_Pi
 - how many CPU cycles are required for a DIV operation?
- STREAM
 - determine memory bandwidth
- Stencil
 - optimization vs. speedup
 - memory access pattern

measuring/getting optimal performance may require process binding