

Performance tools part 2: Performance analysis with hardware metrics

likwid-perfctr



Probing performance behavior

- How do we find out about the performance properties and requirements of a parallel code?
Profiling via advanced tools is often overkill
- A coarse overview is often sufficient: **likwid-perfctr**

Simple end-to-end measurement of hardware performance metrics

Operating modes:

- Wrapper
- Stethoscope
- Timeline
- Marker API

Preconfigured and extensible
metric groups, list with
likwid-perfctr -a



BRANCH: Branch prediction miss rate/ratio
CACHE: Data cache miss rate/ratio
CLOCK: Clock frequency of cores
DATA: Load to store ratio
FLOPS_DP: Double Precision MFlops/s
FLOPS_SP: Single Precision MFlops/s
FLOPS_X87: X87 MFlops/s
L2: L2 cache bandwidth in MBytes/s
L2CACHE: L2 cache miss rate/ratio
L3: L3 cache bandwidth in MBytes/s
L3CACHE: L3 cache miss rate/ratio
MEM: Main memory bandwidth in MBytes/s
TLB: TLB miss rate/ratio
ENERGY: Power and energy consumption

likwid-perfctr wrapper mode

```
$ likwid-perfctr -g L2 -C S1:0-3 ./a.out
```

```
CPU name: Intel(R) Xeon(R) CPU E5-2695 v3 @ 2.30GHz [...]
```

```
<<< PROGRAM OUTPUT >>>
```

```
Group 1: L2
```

Event	Counter	Core 14	Core 15	Core 16	Core 17
INSTR_RETIRED_ANY	FIXC0	1298031144	1965945005	1854182290	1862521357
CPU_CLK_UNHALTED_CORE	FIXC1	2353698512	2894134935	2894645261	2895023739
CPU_CLK_UNHALTED_REF	FIXC2	2057044629	2534405765	2535218217	2535560434
L1D_REPLACEMENT	PMC0	212900444	200544877	200389272	200387671
L2_TRANS_L1D_WB	PMC1	112464863	99931184	99982371	99976697
ICACHE_MISSES	PMC2	21265	26233	12646	12363

```
[... statistics output omitted ...]
```

Metric	Core 14	Core 15	Core 16	Core 17
Runtime (RDTSC) [s]	1.1314	1.1314	1.1314	1.1314
Runtime unhalted [s]	1.0234	1.2583	1.2586	1.2587
Clock [MHz]	2631.6699	2626.4367	2626.0579	2626.0468
CPI	1.8133	1.4721	1.5611	1.5544
L2D load bandwidth [MBytes/s]	12042.7388	11343.8446	11335.0428	11334.9523
L2D load data volume [GBytes]	13.6256	12.8349	12.8249	12.8248
L2D evict bandwidth [MBytes/s]	6361.5883	5652.6192	5655.5146	5655.1937
L2D evict data volume [GBytes]	7.1978	6.3956	6.3989	6.3985
L2 bandwidth [MBytes/s]	18405.5299	16997.9477	16991.2728	16990.8453
L2 data volume [GBytes]	20.8247	19.2321	19.2246	19.2241

Always measured for Intel CPUs

Configured metrics (this group)

Derived metrics

likwid-perfctr stethoscope mode

- likwid-perfctr counts events on cores; it has no notion of what kind of code is running (if any)

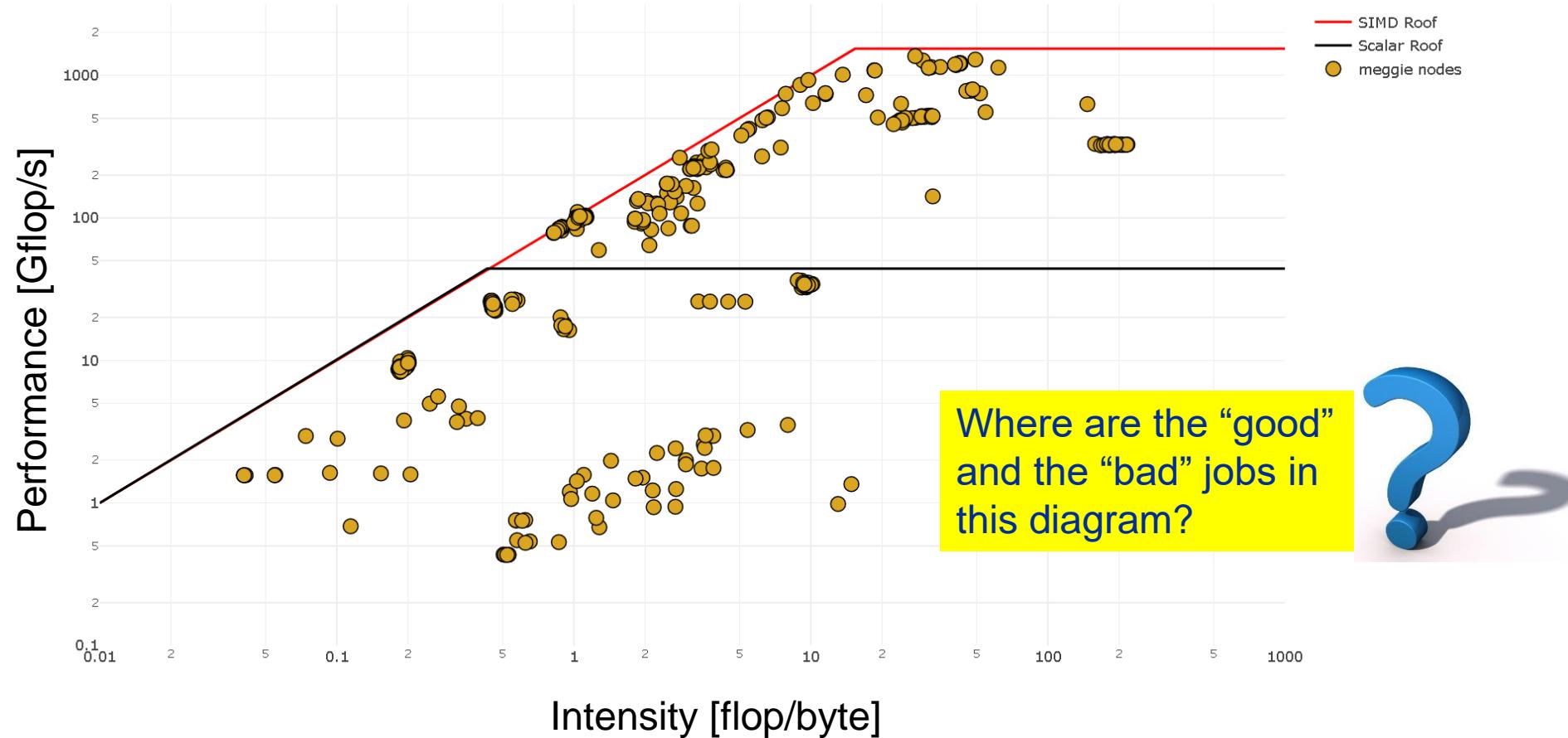
This allows you to “listen” to what is currently happening, **without any overhead**:

```
$ likwid-perfctr -c N:0-11 -g FLOPS_DP -S 10s
```

- It can be used as cluster/server monitoring tool
- A frequent use is to measure a certain part of a long running parallel application from outside

Using Roofline for monitoring “live” jobs on a cluster

Based on measured BW and Flop/s data via likwid-perfctr



<https://github.com/RRZE-HPC/likwid/wiki/Tutorial%3A-Empirical-Roofline-Model>

likwid-perfctr marker API

- The marker API can restrict measurements to code regions
- The API only turns counters on/off. The configuration of the counters is still done by **likwid-perfctr**
- Multiple named regions support, accumulation over multiple calls
- Inclusive and overlapping regions allowed

```
#include <likwid-marker.h>
. . .
LIKWID_MARKER_INIT;                                // must be called from serial region

. . .
LIKWID_MARKER_START("Compute");    // call markers for each thread
. . .
LIKWID_MARKER_STOP("Compute");
. . .
LIKWID_MARKER_START("Postprocess");
. . .
LIKWID_MARKER_STOP("Postprocess");
. . .

LIKWID_MARKER_CLOSE;                                // must be called from serial region
```

Before LIKWID 5
use likwid.h

- Activate macros with **-DLIKWID_PERFMON**
- Run **likwid-perfctr** with **-m** switch to enable marking
- See <https://github.com/RRZE-HPC/likwid/wiki/TutorialMarkerF90> for Fortran example

Compiling, linking, and running with marker API

Compile:

```
cc -I/path/to/likwid.h -DLIKWID_PERFMON -c program.c
```

Link:

```
cc -L/path/to/liblikwid program.o -o program -llikwid
```

Run:

```
likwid-perfctr -C <CPULIST> -g <GROUP> -m ./program
```

- One separate block of output for every marked region
- Caveat: Marker API can cause overhead; do not call too frequently!

So... what should I look at first?

Focus on **resource utilization** and **instruction decomposition!**

Metrics to measure:

- Operation throughput (Flops/s)
- Overall instruction throughput (CPI)
- **Instruction breakdown:**
 - FP instructions
 - loads and stores
 - branch instructions
 - other instructions
- Instruction breakdown to **SIMD width** (scalar, SSE, AVX, AVX512 for x86)
- **Data volumes and bandwidths** to main memory (GB and GB/s)
- Data volumes and bandwidth to different cache levels (GB and GB/s)

Useful diagnostic metrics are:

- Clock frequency (GHz)
- Power (W)

All the above metrics can be acquired using performance groups:

MEM_DP, MEM_SP, BRANCH, DATA, L2, L3

Example: triangular matrix-vector multiplication

```
#define N 10000 // matrix in memory
#define ROUNDS 10
// Initialization
fillMatrix(mat, N*N, M_PI);
fillMatrix(bvec, N, M_PI);

// Calculation loop
#pragma omp parallel
{
    for (int k = 0; k < ROUNDS; k++) {
        #pragma omp for private(current,j)
        for (int i = 0; i < N; i++) {
            current = 0;
            for (int j = i; j < N; j++)
                current += mat[(i*N)+j] * bvec[j];
            cvec[i] = current;
        }
        while (cvec[N>>1] < 0) {dummy();break;}
    }
}
```



Guard with impossible condition and
dummy() function call required if
cvec not used afterwards

Example: triangular matrix-vector multiplication

```
#include <likwid-marker.h>
[...] // defines, fillMatrix, init data
LIKWID_MARKER_INIT;
#pragma omp parallel
{
    for (int k = 0; k < ROUNDS; k++) {
        LIKWID_MARKER_START("Compute");
        #pragma omp for private(current,j)
        for (int i = 0; i < N; i++) {
            current = 0;
            for (int j = i; j < N; j++)
                current += mat[(i*N)+j] * bvec[j];
            cvec[i] = current;
        }
        LIKWID_MARKER_STOP("Compute");
        while (cvec[N>>1] < 0) {dummy();break;}
    }
LIKWID_MARKER_CLOSE;
```



Example: triangular matrix-vector multiplication

```
$ likwid-perfctr -C 0,1,2 -g L2 -m ./a.out
```

```
-----  
CPU type:      Intel Core SandyBridge EN/EP processor  
CPU clock:     3.09 GHz  
-----
```

```
===== Group 1: Region Compute =====
```

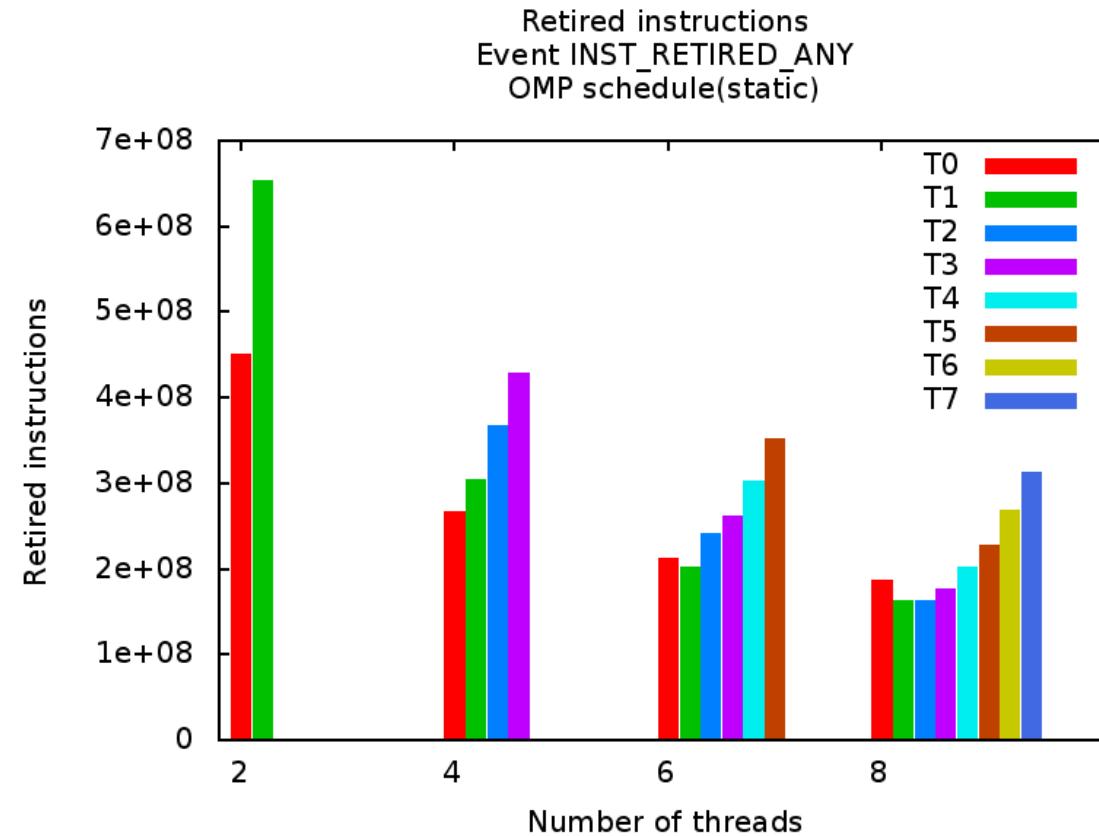
Region Info	Core 0	Core 1	Core 2
RDTSC Runtime [s]	0.161382	0.161365	0.161365
call count	10	10	10
Event	Counter	Core 0	Core 1
INSTR_RETIRED_ANY	FIXC0	2.626800e+08	3.187585e+08
CPU_CLK_UNHALTED_CORE	FIXC1	4.972802e+08	4.961411e+08
CPU_CLK_UNHALTED_REF	FIXC2	4.972801e+08	4.961404e+08
L1D_REPLACEMENT	PMC0	5.490278e+07	3.927353e+07
L1D_M_EVICT	PMC1	2.920200e+04	2.876600e+04
ICACHE_MISSES	PMC2	4.649000e+03	4.984000e+03

???

Example: triangular matrix-vector multiplication

Retired instructions are misleading!

Waiting in implicit OpenMP barrier executes many instructions



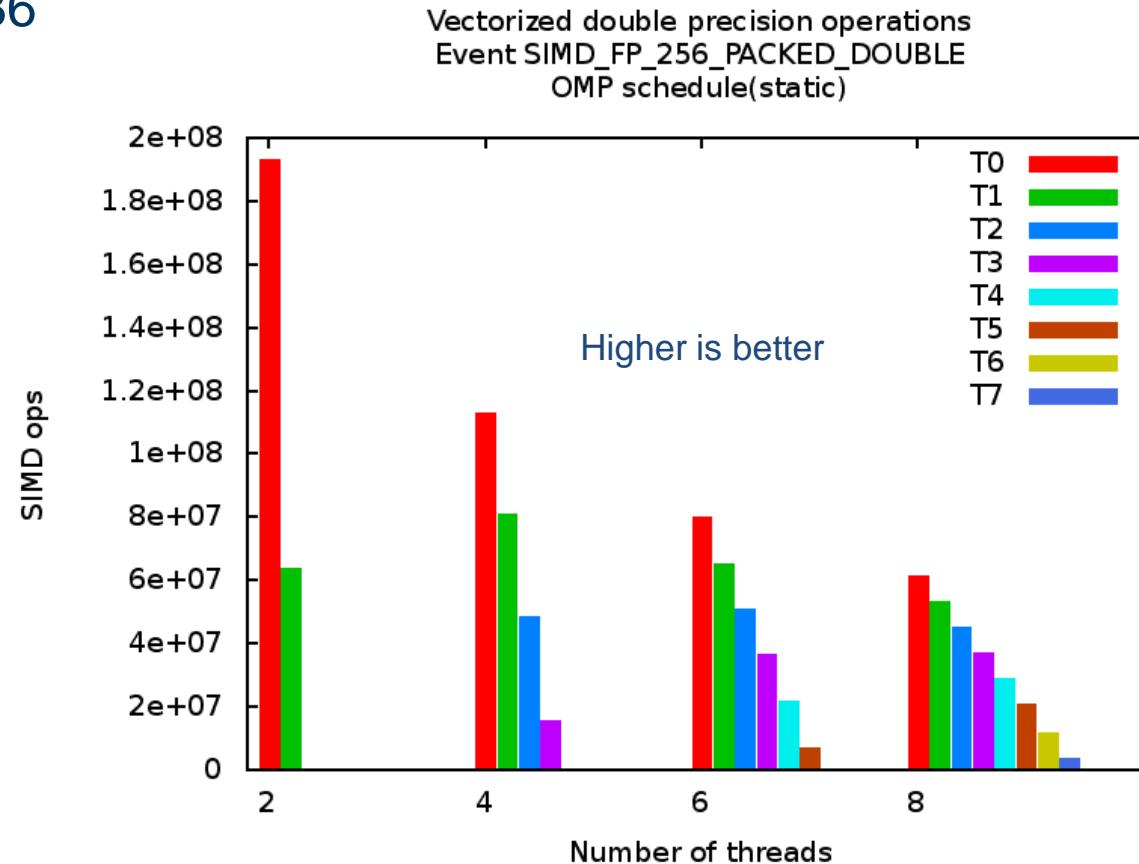
We need to measure actual work (or use a tool that can separate user from runtime lib instructions)

Example: triangular matrix-vector multiplication

Floating-point instructions reliable \leftrightarrow useful work metric

Caveats

- FP instr. counters from SandyBridge to Haswell are only qualitatively correct
- Masked SIMD lanes cannot be counted directly on x86

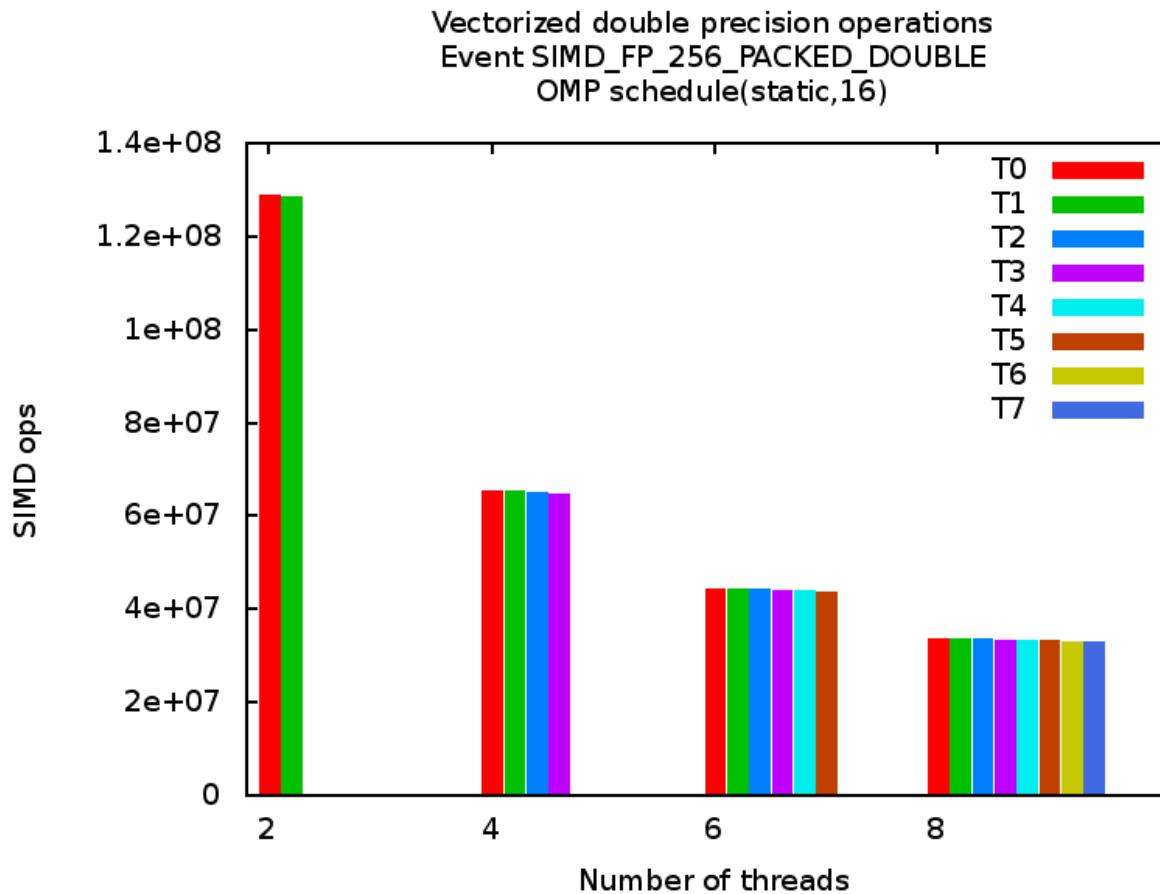


Example: triangular matrix-vector multiplication

Changing OMP schedule to static with chunk size 16 \leftrightarrow smaller work packages per thread

No imbalance anymore!

Is it also faster?



Summary of hardware performance monitoring

- Useful only if you know what you are looking for
 - Hardware event counting bears the potential of acquiring massive amounts of data for nothing!
- Resource-based metrics are most useful
 - Cache lines transferred, work executed, loads/stores, cycles
 - Instructions, CPI, cache misses may be misleading
- Caveat: Processor work != user work
 - Waiting time in libraries (OpenMP, MPI) may cause lots of instructions
 - → distorted application characteristic
- Another very useful application of PM: validating performance models!
 - Roofline is data centric → measure data volume through memory hierarchy