



Analyzing data latency access

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- Data access latency can have a major impact on performance (more data on that topic in the rest of the talk)
- Several hardware mechanisms have been designed to minimize latency impact (latency hiding):
 - Various buffers: Reservation Station, Reorder Buffer, Load/Store Buffer, Line Fill Buffer
 - Hardware prefetchers: on Xeon, 4 prefetchers per core: topic is serious [©]
 - Various memory levels and types (Off chip versus on Chip DRAM on KNL): useful not only for latency but also bandwidth.
 - Special PMU PEBS for tracking load latency...

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Difficult to exactly associate with every load its latency

- Exactly is of major importance because you want to identify the array access which is guilty
- Some PMU events can provide such info.

Even more difficult to associate with each load, its latency impact on the overall loop performance.

- It is one thing to know that you are sick but it is better to know how serious it is and even better to get the right prescriptions
- Accelerate the exploration of the different prefetching configurations
- PMU events have hard time to provide such info.





Various optimization mechanisms:

- Array restructuring, blocking, loop interchange
- Allocating to different memory levels: On chip DRAM versus Off chip DRAM
- Use software instructions to perform aggressive prefetching, more aggressive than hardware due to a better knowledge on loop bounds

BIG ISSUE: how to drive use of these techniques, what is benefit of blocking, prefetching, etc. ???

• PMU do not provide such info on potential gain.

Many simple performance models are only dealing with bandwidth: simplistic with roofline method, more realistic by taking into account different functional units bandwidths and even better different cache levels.





Interesting enough, bandwidth and latency are related through buffer usage, more precisely Little's Law

> Classical formulation: L = λ W

- L : Average Number of customers
- W : Average waiting time
- λ : Average arrival rate



>First use of L = λ W : model off chip cache line access

- L : Average Number of in flight cache line requests (smaller than number of LFB slots: 10!!)
- W : Average waiting time (in cycles) in LFB (close to Latency)
- λ : Average number of cache lines entering per cycle the LFB (close to Bandwidth)

This formula clearly shows that Bandwidth is a function of latency and max number of LFB slots. Clearly constant bandwidth (as assumed by Roof Line model) is a myth ⁽²⁾

Second use: more complex: model FU within CPU.

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UFS: Uop Flow Simulation

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- Ignore semantics: code not executed: use instruction traces. Very simple for loops without branches in loop body.
- Off line analysis/simulation
- Works directly with asm/binary
- Cycle level simulation
- Cycle accurate simulation of core pipeline according to public/published information. If no information is available, use best fit algorithm
- Parametrizable inputs: buffer sizes, instruction latencies/back to back rates, number of ports, bandwidths.....

Hardware parameters: two main options

- 1. Use published numbers: issues with quality and user versus system view
- 2. Design specific benchmark to measure the requested info.

Our approach is with specific benchmarks.

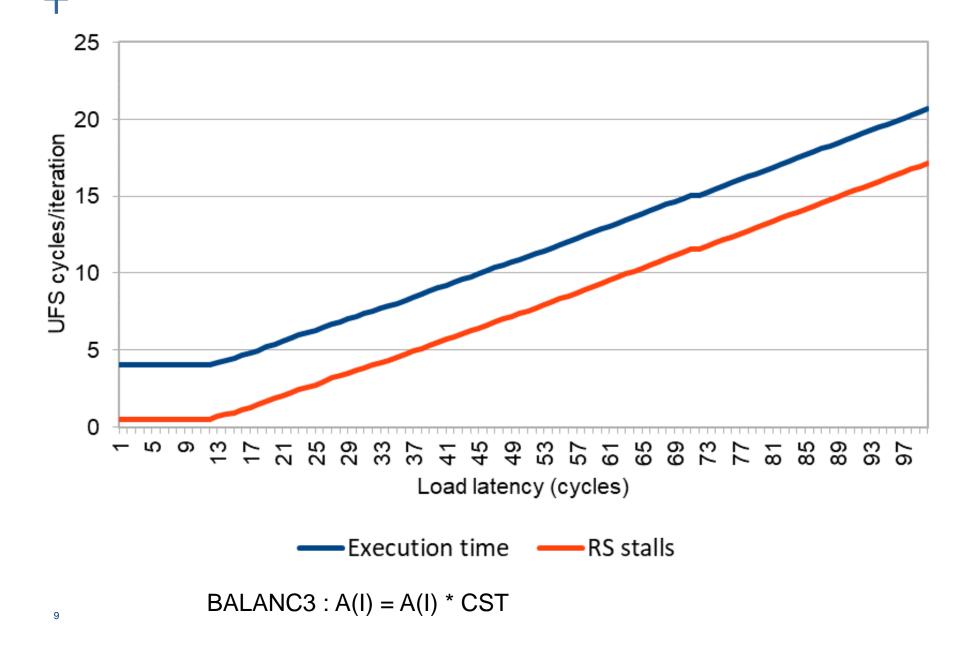


Much faster than most of standard simulators (but does much less)

Why bother with UFS: Yet Another Simulator

- Due to its speed, ability to perform massive parameter studies (Sensitivity Analysis)
- Globally good accuracy most of the time, within 10% of real measurements
- Much better than counters for understanding real issues: the difficulty is that buffer saturation is not the source of the problem but in general the consequence of a problem.... Excessive load latencies will quickly lead to Load Buffer and/or RS overflow. However, well controlled latencies will avoid such overflow and lead to better performance.
- ONE MAJOR ISSUE: how to optimize code to minimize RS, ROB etc...footprint/consumption





Exascale Characterizing codelet response to latency variations



On most of the latency sensitivity plots (see previous slides), two regions can be clearly seen:

- Initial flat region where codelet is supporting higher latencies without any impact on performance (buffers filling up).
- Linear region where impact on performance varies linearly with latency increase (one of the buffer is full).

Two key parameters for characterizing a codelet:
Lmax: maximum latency tolerated with 0 or negligible cost
Additional cost: slope of the linear line

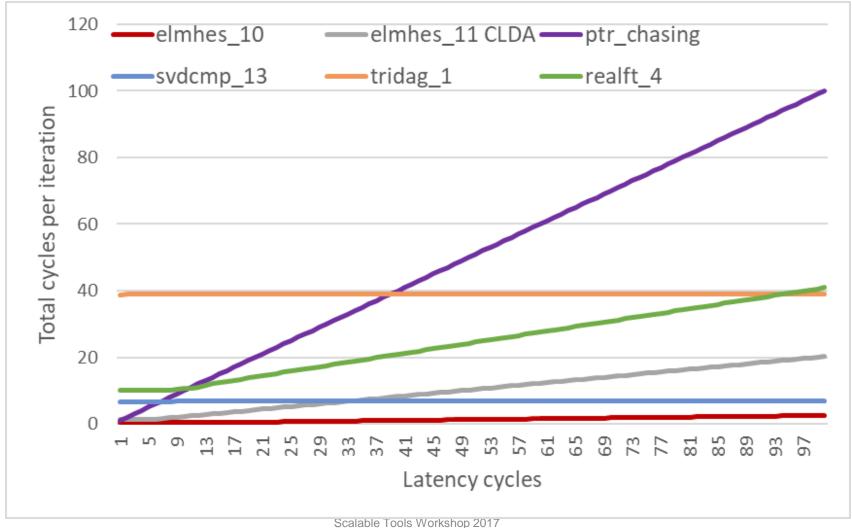


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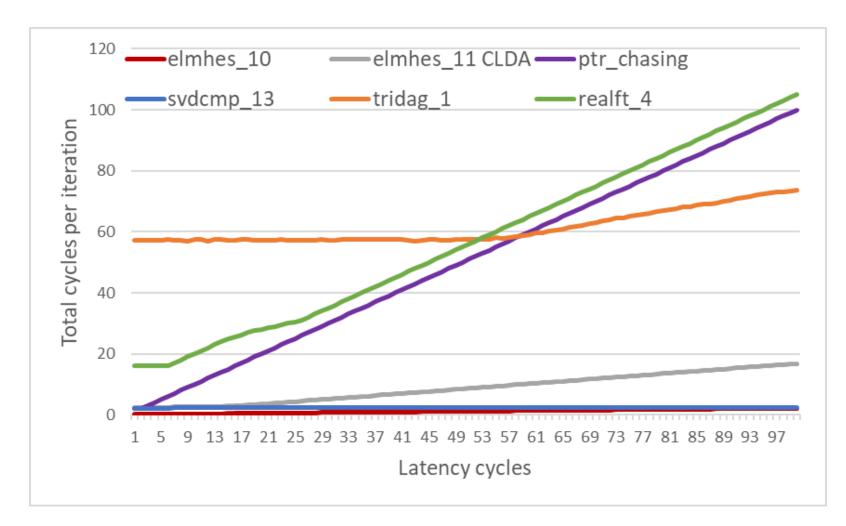
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For KNL







UFS allows to model the impact of varying latency: this can be done uniformly on loads and stores or individually.

This allows to understand the potential performance gain of:

- 1. Better blocking (blocking for L2 instead of L3)
- 2. Better prefetching (add extra prefetch instructions on targeted loads)
- 3. Using on die DRAM versus external DRAM (cf. KNL)



- Various Out of Order buffers (ROB, RS, PRF, LB, SB) are critical to get peak performance on modern cores.
- UFS gives a detailed insight on buffer usage and can correlate usage with code.
- > UFS is fast, allowing massive parameter explorations.
- UFS is excellent at exploring what if scenarii (hardware and software)
- UFS allows to characterize latency impact on loop performance.

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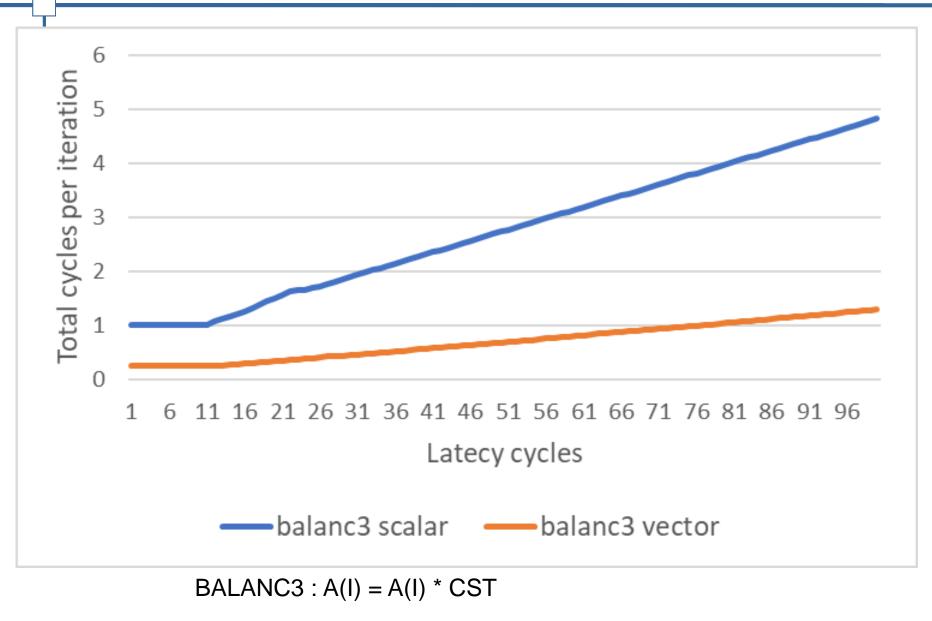




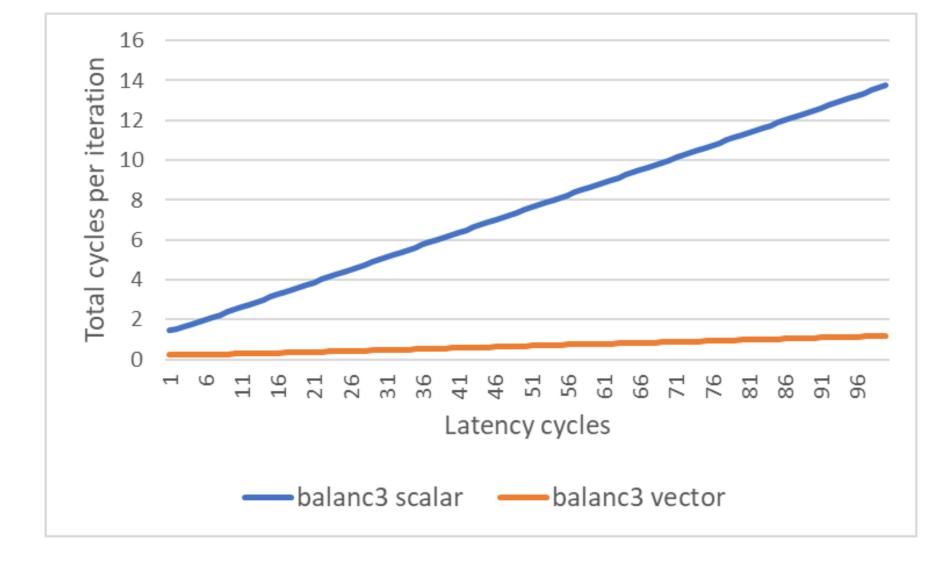
BACKUP SLIDES

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CQA: Code Quality Analyzer Open Source: www. maqao.org STATIC MODEL: all operands are assumed resident in L1. Compute 3 bounds:

- Issue/Decode : divide number of uops per 4 + ceiling effect
- Execution: count number of instructions per port/FU (taking into account rate)
- Inter iterations dependencies: compute cycles

Predicted number of cycles = max of the 3 estimates above.

THROUGHPUT/BANDWIDTH BASED MODEL



Analyze ASM:

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12 FP Mul instructions, 16 FP Add/Sub instructions, 4 Loads Instructions + 4 Address computations, 4 Store Instructions, 18 Alu Instructions

Compute the 3 bounds:

- 1. Issue/Decode : 58 uops after unlamination.: 14,5 cycles rounded to 15 cycles.
- Execution: P0 (FP */ALU): 15 cycles, P1 (FP+/ALU): 16 cycles, P2 (Load): 4 cycles, P3 (Load): 4 cycles, P4 (Store): 4 cycles, P5 (Misc/ALU): 15 cycles
- 3. Inter iterations dependencies: 13 cycles

Predicted number of cycles = max of the 3 estimates = 16 cycles



CQA prediction: 16 cycles Measurement: 23,36 cycles GAP: 23,36-16 = 7,36 cycles. BEYOND L1: results slightly worse but to be expected

What happens??

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A first answer provided by hardware events: each of the buffer has an associated event counting the number of cycles where when full it causes the front end to stall.

Measurement: Reservation Station Stalls occur for 7,85 cycles....

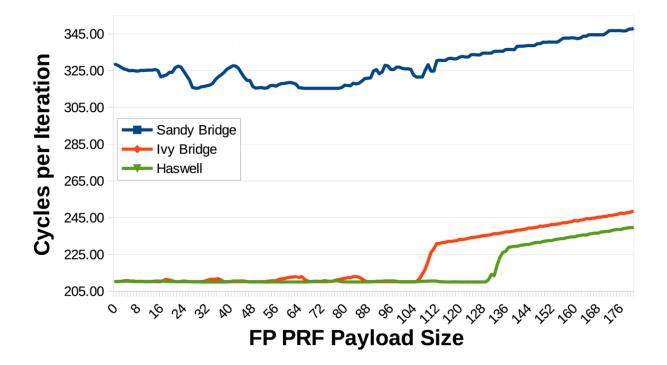
ISSUES: a stall at the front end does not necessarily result in cycles lost in the back end, multiple counting (several buffers full)

UFS results:

23,01 (using SNB buffer sizes): perfect match with measurements and points to RS full leading to wasted cycles

19,03 (using large buffers): time lost in dispatch





PRINCIPLE: increase the payload to force an overflow in the target resource which in turn translates into a discontinuity in timing.

"payload" is basically some extra instructions specifically designed to run COMPLETELY in parallel with divisions *UNLESS* they saturate the target buffer.

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