Assessing Memory Bandwidth on ARCHER2 and LUMI Using CAMP

CUG 11.5.23, Helsinki

Wenqing Peng, Adrian Jackson and Evgenij Belikov* EPCC, The University of Edinburgh



* corresponding author: e.belikov@epcc.ed.ac.uk

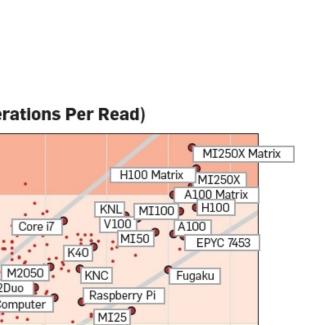
Outline

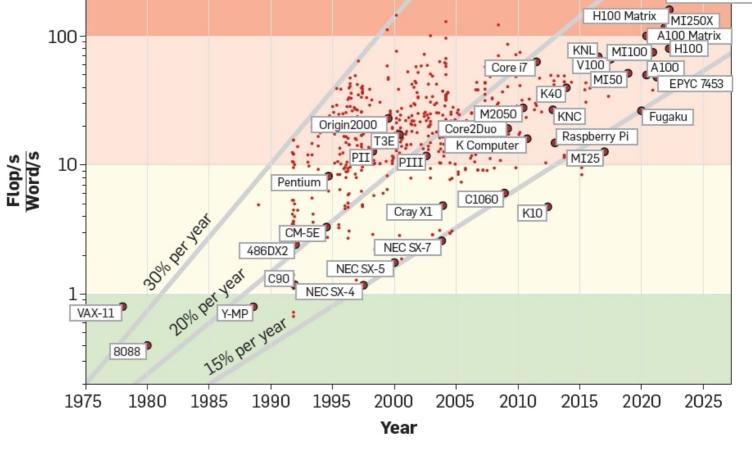
- Motivation
- CAMP Overview
- CAMP custom kernel example (dot product)
- Experimental Setup
- Results on ARCHER2, LUMI and NEXTGenIO
- Conclusions and Future Work
- Discussion

Machine Imbalance

- Memory Wall and growing machine imbalance
- NUMA architectures with increasing number of cores
- Many apps that were considered computebound now memorybound

Machine Balance (Double-Precision Floating Point Operations Per Read)

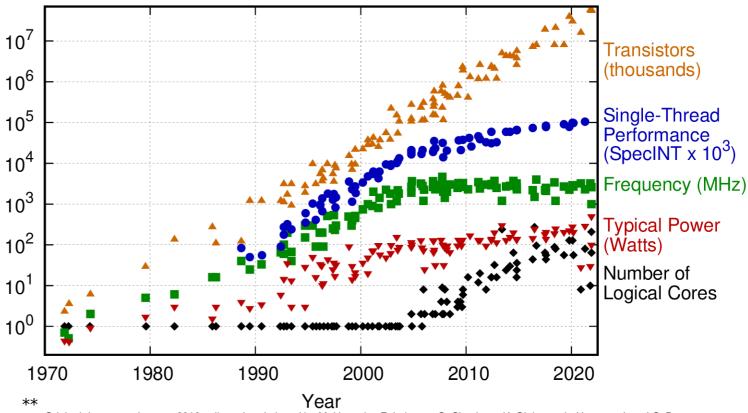




Deep Memory Hierarchies

- Deeper memory hierarchies to accommodate larger core counts
- Memory access: one of the main intra-node bottlenecks
- Various access patterns and operational intensities in kernels → CAMP

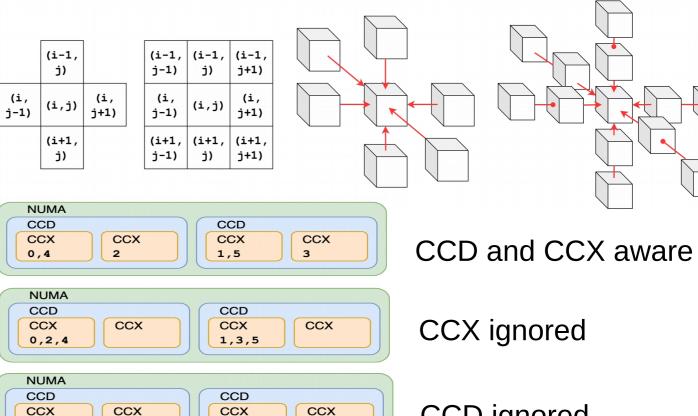
50 Years of Microprocessor Trend Data

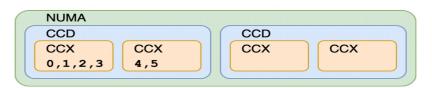


Original data up to the year 2010 collected and plotted by M. Horowitz, F. Labonte, O. Shacham, K. Olukotun, L. Hammond, and C. Batten New plot and data collected for 2010-2021 by K. Rupp

Introduction to CAMP (Configurable App for Memory Probing)

- Operational intensity (OI)
- Different access patterns reflecting *spatial* and *temporal locality*
- Thread pinning _____
- Custom kernels
- Performance Counters
- Portability





1,3,5

0,2,4

CCD ignored

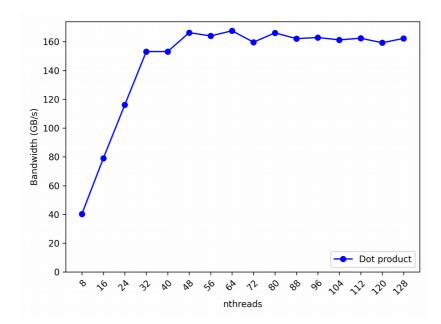
CCD and CCX ignored

* for more details see:

CAMP: a Synthetic Micro-Benchmark for Assessing Deep Memory Hierarchies. HiPar@SC'22: 28-36

CAMP: Extensions

- More fine-grained OI (from 0.08 to 0.02)
- Energy consumption analysis
- Support for custom kernels: dot product
- New way to configure runs via a config file and using an outer-level driver script



Algorithm 1 CAMP custom kernel (pseudocode)

procedure CAMP_PREPROCESS

#pragma omp parallel default(none) shared(a, b) {
 int tid = omp_get_thread_num();
 a[tid] = malloc(size);
 b[tid] = malloc(size);
 for i in range(1, size) do
 a[tid][i] = 1.0;
 b[tid][i] = 2.0;
 end for
 }
end procedure

```
procedure CAMP_KERNEL
    #pragma omp parallel default(none) shared(a, b) reduc-
tion(+:sum) {
    int tid = omp_get_thread_num();
    for i in range(1, size) do
        sum += a[tid][i] * b[tid][i];
    end for
    }
end procedure
procedure CAMP_POSTPROCESS
    #pragma omp parallel default(none) shared(a, b) {
    int tid = omp_get_thread_num();
    free(a[tid]);
    free(b[tid]);
    }
end procedure
```

Experimental Setup

- ARCHER2: AMD Rome, node hierarchy on the right*
- LUMI: AMD Milan
- NEXTGenIO: Intel Ice Lake
- CAMP cyclic distribution (similar to OpenMP spread)
- data size >> L3 cache size
- optimisation on, vectorisation on
- best of 10 runs (raw data kept)

NUMA 2

eocc

CCD

CCD

хGМ

xGM

IF-Switch

DDR4 UMC DDR4 UMC GMI2 GMI2 GMI2 I/O Die I/O Die

NUMA 0

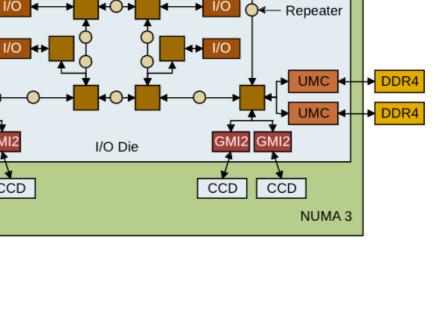
NUMA 1

CCD

CCD

xG№

xGM



Systems Overview: ARCHER2, LUMI-C, NEXTGenIO-icx

ARCHER2

LUMI-C

NEXTGenIO-icx

Intel Ice Lake

56 (2x28)

2.00

1280

48

42

256

Xeon Gold 6330

Hardware

Architecture	AMD Rome	AMD Milan
Model	EPYC 7742	EPYC 7763
Cores per node	128 (2x64)	128 (2x64)
CPU freq. (GHz)	2.25	2.45
L1\$ (KB)	32	32
L2\$ (KB)	512	512
L3\$ (MB)	16x16	8x32
RAM (GB)	256	256

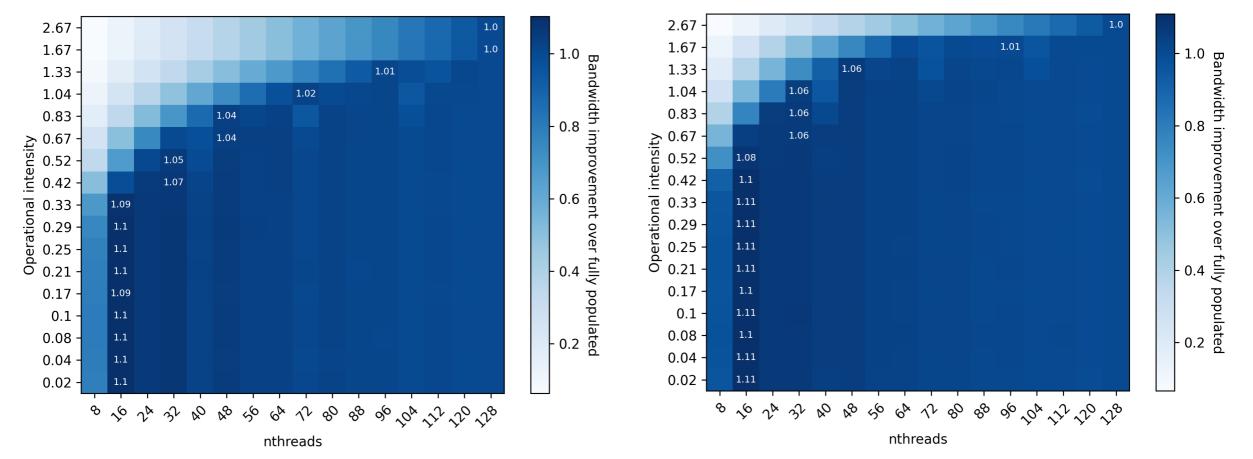
Software

O S	SLES 15	SLES 15	CentOS 7.9
Compiler	PrgEnv-cray	PrgEnv-cray	Intel oneAPI 2023
Python	3.8.5	3.9.13	3.8.5
numpy	1.24.2	1.20.0	1.24.2
pandas	1.5.3	1.3.4	1.5.3
matplotlib	3.7.1	3.7.1	3.7.1

Node		
Socket		
NUMA		
CCD	CCD	
CCX 0,1,2,3 CCX 4,5,6,7	CCX 8,9,10,11 CCX 12,13,14,15	
3 more NUMA nodes		
Socket		
NUMA		
CCD	CCD	
CCX CCX	CCX CCX	
64,65,66,67 68,69,70,71	72,73,74,75 72,73,74,75	
3 more NUMA nodes		

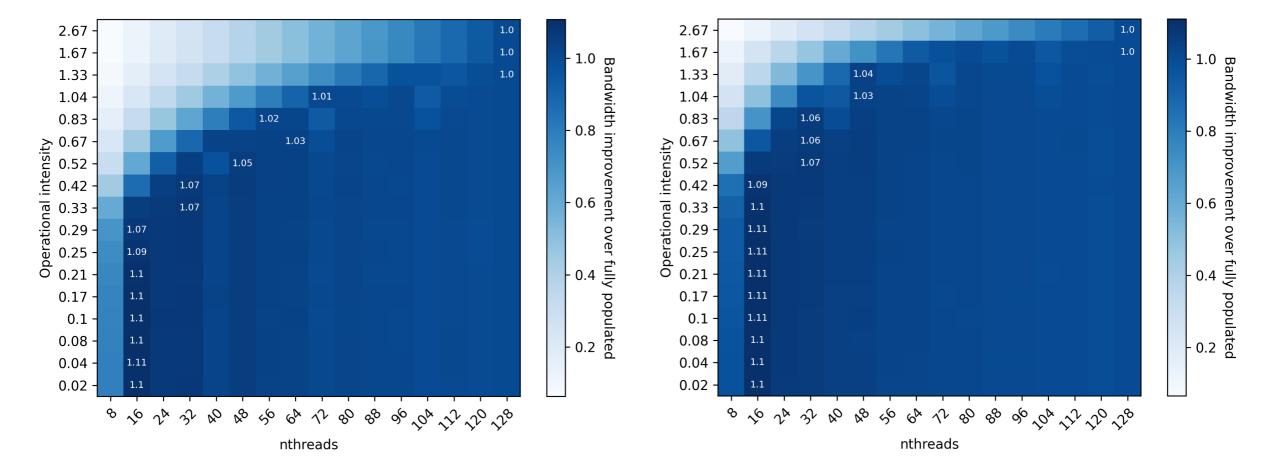
ARCHER2 node hierarchy

ARCHER2 Results: Contiguous (1.50 GHz vs 2.25 GHz CPU freq.)



 Highest bandwidth in most cases with 16 threads (up to 11% increase); stronger effect for higher CPU freq.

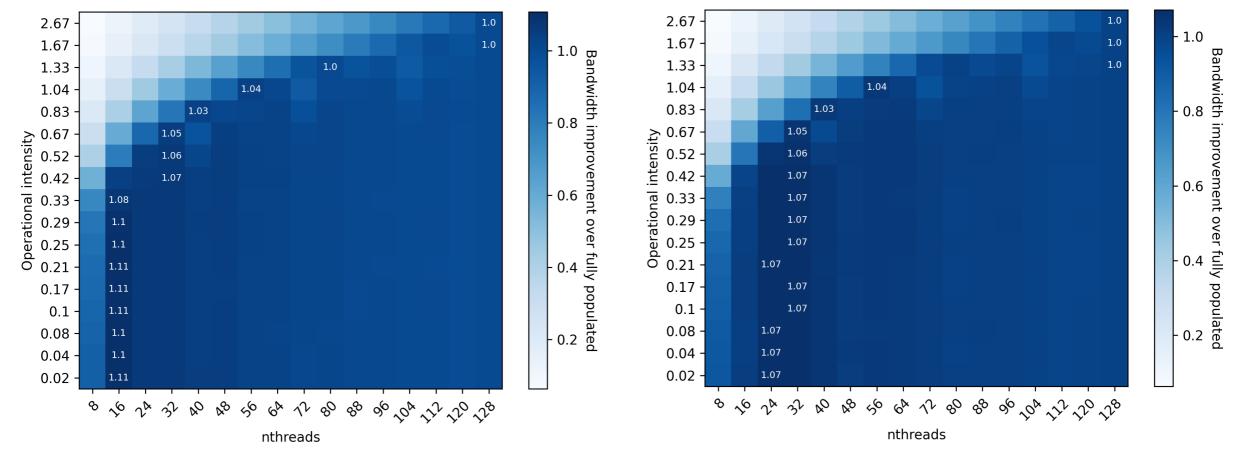
ARCHER2 Results: Stride4 (1.50 GHz vs 2.25 GHz CPU freq.)



 Highest bandwidth in most cases with 16 threads (up to 11% increase); stronger effect for higher CPU freq.

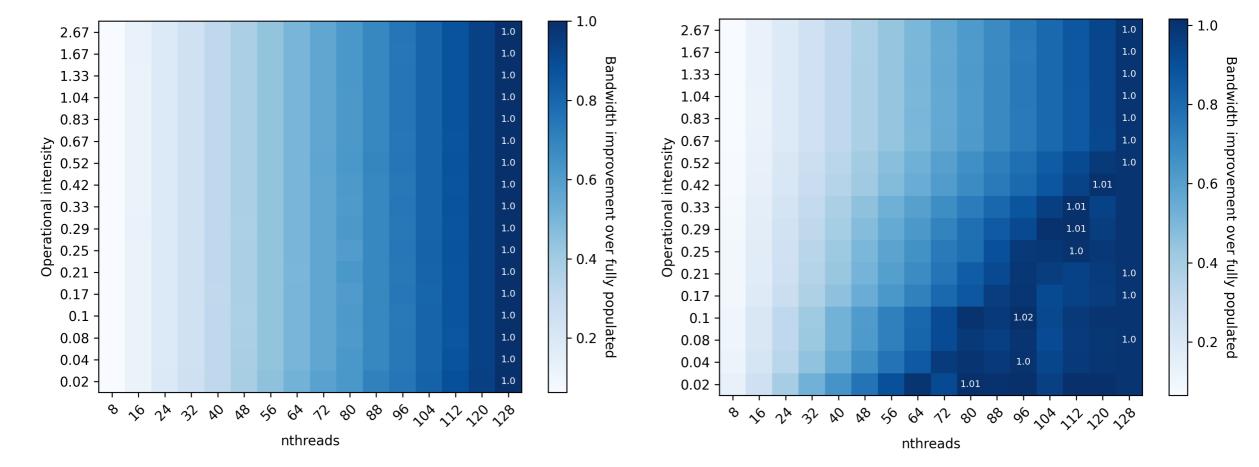
10

Effect of sub-optimal hierarchy setting (ARCHER2 Stride4 2 GHz)



 Need to take case when specifying thread placement at sub-NUMA region level (NB: poor settings on the right)

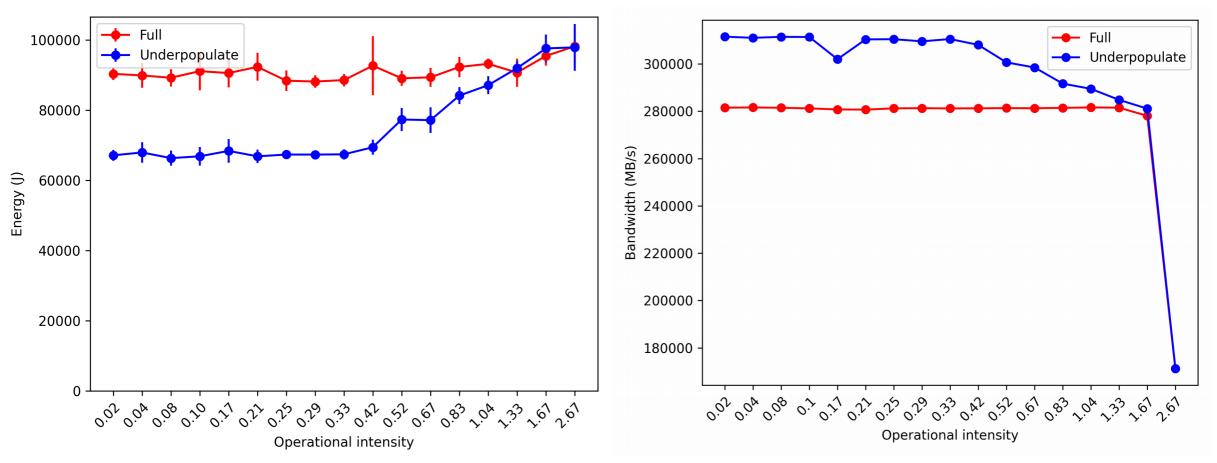
ARCHER2 Results: Stencil5 (1.50 GHz vs 2.25 GHz CPU freq.)



Effect only visible for higher CPU freq.
 (e.g. +2% using 96 threads for operational intensity of 0.1)

12

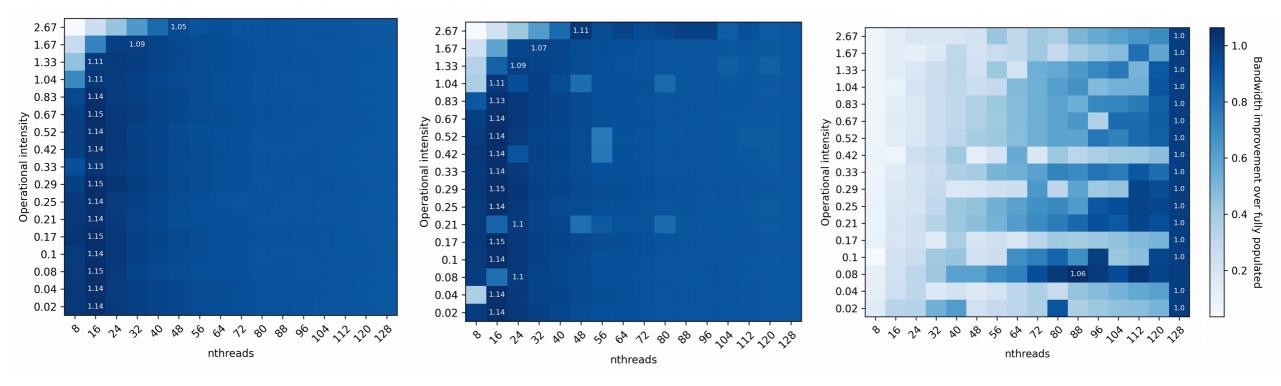
ARCHER2: Energy Consumption and Absolute Bandwidth



Contig. access: underpopulated case has lower energy cost (r=1000)

Absolute bandwidth higher for underpopulated setup

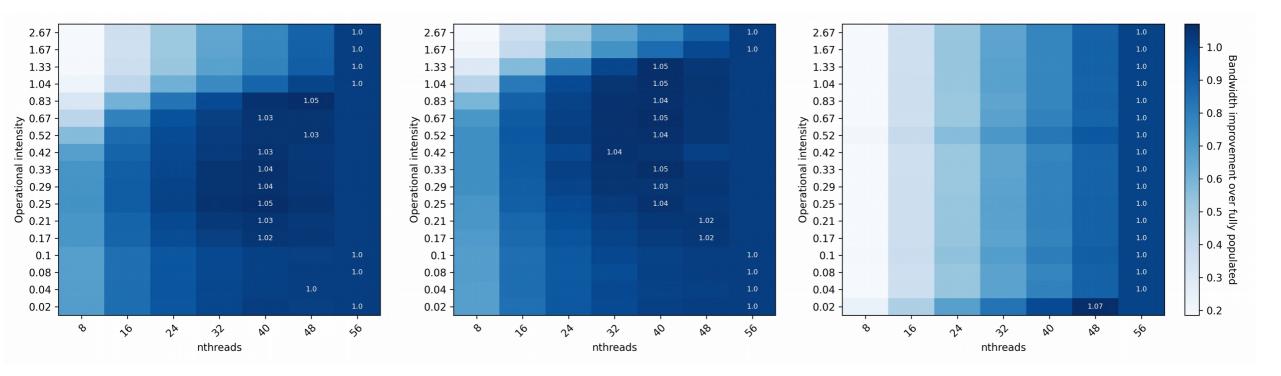
LUMI Results: Contiguous, Stride4, Stencil5 (default CPU freq)



- Highest bandwidth in most cases with 16 threads (up to +15%) for contiguous and strided access patterns
- Unclear patchy behaviour for stencil (for <128 threads)

14

NEXTGenIO Results: Contiguous, Stride4, Stencil5



- Highest bandwidth in most cases with 40 threads (up to +5%) for contiguous and strided access patterns
- For stencil +7% only for smallest operational intensity

Conclusions

• CAMP is an open-source tool complementing existing benchmarks with focus on assessing intra-node memory hierarchy and support for different access patterns with tunable operational intensity

→ https://github.com/CAMP-benchmark/CAMP

- We observe underpopulation effects (up to 11% and 15% increased bandwidth on ARCHER2 and LUMI-C respectively, when using only 16 threads rather than 128); the effect appears stronger for higher CPU frequency settings
- Using lower CPU frequency setting on ARCHER2 results in increased runtime and energy consumption
- Underpopulation may help reduce energy consumption

Future Work

- Next we plan to extend CAMP by adding GPU support
- Increase the number of supported access patterns
- Grow the library of kernels and compare to efficient implementations (e.g. BLAS, FFT)
- Apply CAMP to cost modelling for applications

17

epc

Acknowlegments

This work used the ARCHER2 UK National Supercomputing Service (https://www.archer2.ac.uk), the NEXTGenIO system (http://www.nextgenio.eu) and LUMI (https://www.lumisupercomputer.eu).

The NEXTGenIO system was funded by the European Union's Horizon 2020 Research and Innovation program under Grant Agreement no. 671951, and supported by EPCC, The University of Edinburgh.

We acknowledge the EuroHPC Joint Undertaking for awarding this project access to the EuroHPC supercomputer LUMI, hosted by CSC (Finland) and the LUMI consortium through a EuroHPC Regular Access call.

Some absolute numbers

