

Collective Memory Transfers for Multi-Core Chips

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- Future technologies will allow more parallelism on chip
- Computational throughput expected to increase faster than memory bandwidth
 - Pin and power limitations for memory
- Many applications are limited by memory bandwidth
- We propose a mechanism to coordinate memory accesses between numerous processors such that the memory is presented with in-order requests
 - Increases DRAM performance and power efficiency

Today's Menu



- Today's and future challenges
- The problem
- Collective memory transfers
- Evaluation
- Related work, future directions and conclusion

Chip Multiprocessor Scaling



By 2020 we may witness 2048-core chip multiprocessors

Intel 80-core



NVIDIA Fermi: 512 cores



AMD Fusion: four full CPUs and 408 graphics cores



How to stop interconnects from hindering the future of computing. OIC 2013

Straw-man Exascale Processor

RF (512)

DP FP

FMAC

I\$ (16KB)

Data

(64KB)

Application specific



8 Exe + Control





Processor Chip (~16 Clusters)

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Technology	4nm, 2020
Die area	16x16 mm2
Cores/die	2000
Frequency	1.1 GHz@Vdd
TFLOPs	4 TF Peak@Vdd
Power	15 W@Vdd
E Efficiency	4 pJ/F@Vdd, much better at NTV

Data Movement and Memory Dominate

rrrrr

lmi



Memory Bandwidth a Constraint





Exascale computing technology challenges. VECPAR 2010



- Parallelism will increase
- Compute capacity increases faster than memory bandwidth
 - 10% memory bandwidth increase per year [1]
 - Compute capacity increase driven by Moore's law
- Data movement and memory access power already a limiting factor
 - Projected to worsen with future technologies
- Numerous applications are memory bandwidth bound
 - Will become worse in the future

[1] Scaling the bandwidth wall: challenges in and avenues for CMP scaling. ISCA 2009

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Computation on Large Data





Full 3D Generalization





Data-Parallelism Covers a Broad Range of Applications



- From HPC to embedded computing
- Data-parallel applications a major driver for multi-cores



Convergence of recognition, mining, and synthesis workloads and its implications. Proc. IEEE 2008

The Problem: Unpredictable and Random Order Memory Access Pattern





This is a DRAM Array





Random Order Access Patterns Hurt DRAM Performance and Power



Reading tile 1 requires row activation and copying

	Tile line 1	Tile line 2	Tile line 3
	Tile line 4	Tile line 5	Tile line 6
In order requests: 3 activations	Tile line 7	Tile line 8	Tile line 9
Morst case:			
In order requests: 3 activations Worst case:	Tile line 7	Tile line 8	Tile line 9

9 activations

Impact



- DRAMSim2 [2] with simple in-order and out-of-order traces
 - A single request accesses one 64-Byte word
 - FRFCFS memory scheduler
 - 16MB DDR3 Micron memory module
- DRAM throughput drops 25% for loads and 41% for stores
- Median latency increases 23% for loads and 64% for stores
- Power increases by 2.2x for loads and 50% for stores

[2] DRAMSim2: A cycle accurate memory system simulator. IEEE CAL 2011

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Collective Memory Transfers

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Hierarchical Tiled Arrays to Transfer Data Layout Information





"The hierarchically tiled arrays programming approach". LCR 2004

Hierarchical Tiled Arrays to Transfer Data Layout Information



Array = hta(name, {[1,3,5],[1,3,5]}, [3,3], **F(x) = x**); // Mapping function or matrix

Loading a HTA with a CMS read

HTA_instance = CMS_read (HTA_instance);

Loading the same HTA with DMA operations for each line of data

Array[row1] = DMA (Starting_address_row1, Ending_address_row1);

Array[rowN] = DMA (Starting_address_rowN, Ending_address_rowN);

"The hierarchically tiled arrays programming approach". LCR 2004



- If data array is not tiled, transferring the layout information over the on-chip network is too expensive
- Instead, the CMS engine learns the mapping by observing each processor's requests in the first iteration of the application's loop



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Execution Time Impact





- Up to 55% application execution time reduction due to memory b/w
 - 27% geometric mean

Execution Time Impact





- 31% improvement for dense grid applications. 55% for sparse
- Sparse grid applications have lower computation times therefore they exert more pressure to the memory

Relieving Network Congestion







CMS significantly simplifies the memory controller because shorter FIFO-only transaction queues are adequate

ASIC Synthesis	DMA	CMS
Combinational area (µm ²)	743	16231
Non-combinational area (µm ²)	419	61313
Minimum cycle time (ns)	0.6	0.75

To offset the cycle time increase, we can add a pipeline stage (insignificant effect compared to the duration of a transaction)

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- A plethora of memory controller schedulers
 - However, the majority are passive policies that do not control the order requests arrive to the memory controller
 - Can only choose from within the transaction queue
- LLCs can partially re-order writes to memory (if write-back)
 - Write-through caches preferable in data-parallel computations [3]
 - CMS focuses on fetching new data and writing old data
- Prefetching focuses on latency, not bandwidth
 - Mispredictions are possible
 - Lacks application knowledge
- Past work uses injection control [4] or routers to partially re-order requests [5]
 - [3] Stencil computation optimization and auto-tuning on state-of-the-art multicore architectures. SC 2008
 - [4] Entry control in network-on-chip for memory power reduction. ISLPED 2008
 - [5] Complexity effective memory access scheduling for many-core accelerator architectures. MICRO 2009



- What is the best interface to CMS from the software?
 - A library with an API similar to DMA function calls (the one shown)?
 - Left to the compiler to recognize collective transfers?
- How would this work with hardware-managed cache coherency?
 - Prefetchers may need to recognize and initiate collective transfers
 - Collective prefetching?
 - How to modify MESI to support force-feeding data to L1s

Conclusions



- Memory bandwidth will be an increasing limiting factor in application performance
- We propose a software-hardware collective memory transfer mechanism to present the DRAM with in-order accesses
 - Cores access the DRAM as a group instead of individually
- Up to 55% application execution time decrease
 - 27% geometric mean

Questions?



