

MVAPICH2-GDR: Pushing the Frontier of HPC and Deep Learning

Talk at NVIDIA booth (SC 2017)

by

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Outline

- Overview of the MVAPICH2 Project
- MVAPICH2-GPU with GPUDirect-RDMA (GDR)
- What's new with MVAPICH2-GDR
 - Multi-stream Communication for IPC
 - CMA-based Intra-node Communication Support
 - Maximal overlap in MPI Datatype Processing
 - Initial support for GPUDirect Async feature
- Streaming Support with IB Multicast and GDR
- High-Performance Deep Learning (HiDL) with MVAPICH2-GDR
- Conclusions

Overview of the MVAPICH2 Project

- High Performance open-source MPI Library for InfiniBand, Omni-Path, Ethernet/iWARP, and RDMA over Converged Ethernet (RoCE)
 - MVAPICH (MPI-1), MVAPICH2 (MPI-2.2 and MPI-3.0), Started in 2001, First version available in 2002
 - MVAPICH2-X (MPI + PGAS), Available since 2011
 - Support for GPGPUs (MVAPICH2-GDR) and MIC (MVAPICH2-MIC), Available since 2014
 - Support for Virtualization (MVAPICH2-Virt), Available since 2015
 - Support for Energy-Awareness (MVAPICH2-EA), Available since 2015
 - Support for InfiniBand Network Analysis and Monitoring (OSU INAM) since 2015
 - Used by more than 2,825 organizations in 85 countries
 - More than 433,000 (> 0.4 million) downloads from the OSU site directly
 - Empowering many TOP500 clusters (June '17 ranking)
 - 1st, 10,649,600-core (Sunway TaihuLight) at National Supercomputing Center in Wuxi, China
 - 15th, 241,108-core (Pleiades) at NASA
 - 20th, 462,462-core (Stampede) at TACC
 - 44th, 74,520-core (Tsubame 2.5) at Tokyo Institute of Technology
 - Available with software stacks of many vendors and Linux Distros (RedHat and SuSE)
 - <u>http://mvapich.cse.ohio-state.edu</u>
- Empowering Top500 systems for over a decade
 - System-X from Virginia Tech (3rd in Nov 2003, 2,200 processors, 12.25 TFlops) ->
 - Sunway TaihuLight (1st in Jun'17, 10M cores, 100 PFlops)

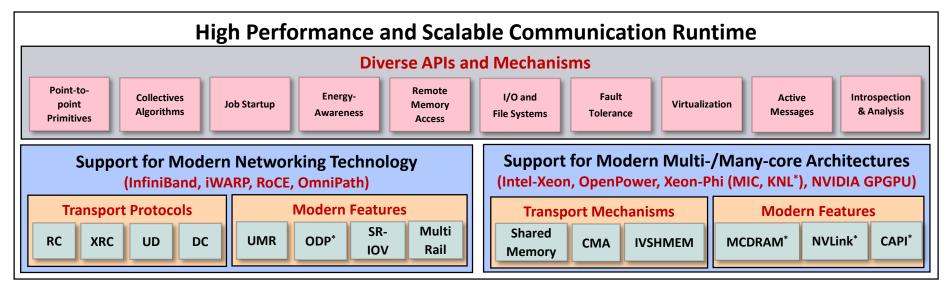
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10 Years & Going Strong!

MVAPICH2 Architecture

High Per	High Performance Parallel Programming Models		
Message Passing Interface	PGAS	Hybrid MPI + X	
(MPI)	(UPC, OpenSHMEM, CAF, UPC++)	(MPI + PGAS + OpenMP/Cilk)	



Upcoming

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MVAPICH2 Software Family

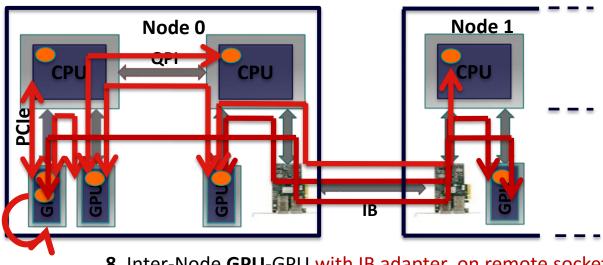
	High-Performance Pa	gh-Performance Parallel Programming Libraries	
	MVAPICH2	Support for InfiniBand, Omni-Path, Ethernet/iWARP, and RoCE	
	MVAPICH2-X	Advanced MPI features, OSU INAM, PGAS (OpenSHMEM, UPC, UPC++, and CAF), an MPI+PGAS programming models with unified communication runtime	
	MVAPICH2-GDR	Optimized MPI for clusters with NVIDIA GPUs	
	MVAPICH2-Virt	High-performance and scalable MPI for hypervisor and container based HPC cloud	
	MVAPICH2-EA	Energy aware and High-performance MPI	
	MVAPICH2-MIC	Optimized MPI for clusters with Intel KNC	
Microbenchmarks			
	ОМВ	Microbenchmarks suite to evaluate MPI and PGAS (OpenSHMEM, UPC, and UPC++) libraries for CPUs and GPUs	

0 C, and UPC++) Tools **OSU INAM** Network monitoring, profiling, and analysis for clusters with MPI and scheduler integration Utility to measure the energy consumption of MPI applications **OEMT**

, and CAF), and

MVAPICH2-GDR: Optimizing MPI Data Movement on GPU Clusters

• Connected as PCIe devices – Flexibility but Complexity



Memory buffers

- 1. Intra-GPU
- 2. Intra-Socket GPU-GPU
- **3**. Inter-Socket **GPU**-GPU
- 4. Inter-Node GPU-GPU
- 5. Intra-Socket GPU-Host
- 6. Inter-Socket GPU-Host
- 7. Inter-Node GPU-Host

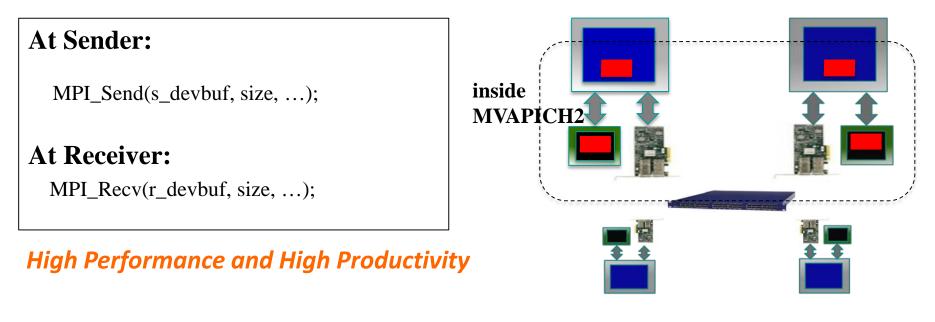
8. Inter-Node GPU-GPU with IB adapter on remote socket

and more . . .

- For each path different schemes: Shared_mem, IPC, GPUDirect RDMA, pipeline ...
- Critical for runtimes to optimize data movement while hiding the complexity

GPU-Aware (CUDA-Aware) MPI Library: MVAPICH2-GPU

- Standard MPI interfaces used for unified data movement
- Takes advantage of Unified Virtual Addressing (>= CUDA 4.0)
- Overlaps data movement from GPU with RDMA transfers



CUDA-Aware MPI: MVAPICH2-GDR 1.8-2.3 Releases

- Support for MPI communication from NVIDIA GPU device memory
- High performance RDMA-based inter-node point-to-point communication (GPU-GPU, GPU-Host and Host-GPU)
- High performance intra-node point-to-point communication for multi-GPU adapters/node (GPU-GPU, GPU-Host and Host-GPU)
- Taking advantage of CUDA IPC (available since CUDA 4.1) in intra-node communication for multiple GPU adapters/node
- Optimized and tuned collectives for GPU device buffers
- MPI datatype support for point-to-point and collective communication from GPU device buffers
- Unified memory

Using MVAPICH2-GPUDirect Version

• MVAPICH2-2.3 with GDR support can be downloaded from

https://mvapich.cse.ohio-state.edu/download/mvapich2gdr/

- System software requirements
 - Mellanox OFED 3.2 or later
 - NVIDIA Driver 367.48 or later
 - NVIDIA CUDA Toolkit 7.5/8.0/9.0 or later
 - Plugin for GPUDirect RDMA

http://www.mellanox.com/page/products dyn?product family=116

- Strongly recommended
- GDRCOPY module from NVIDIA

https://github.com/NVIDIA/gdrcopy

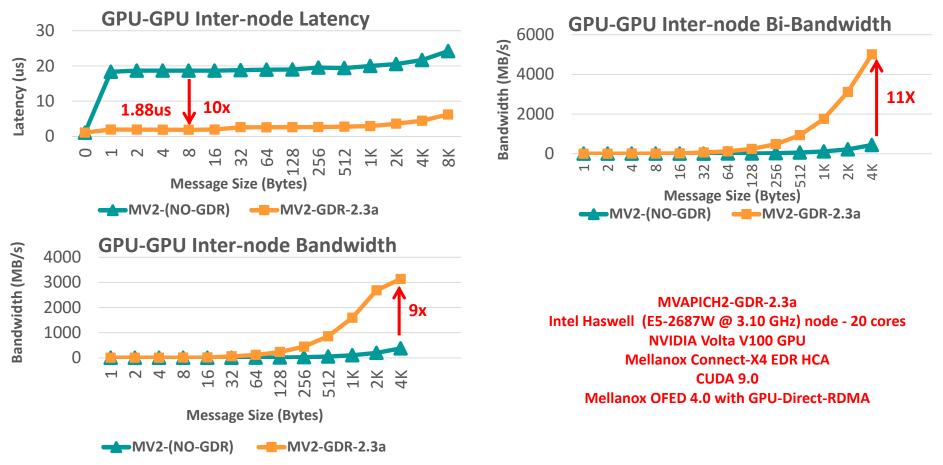
• Contact MVAPICH help list with any questions related to the package

mvapich-help@cse.ohio-state.edu

MVAPICH2-GDR 2.3a

- Released on 11/09/2017
- Major Features and Enhancements
 - Based on MVAPICH2 2.2
 - Support for CUDA 9.0
 - Add support for Volta (V100) GPU
 - Support for OpenPOWER with NVLink
 - Efficient Multiple CUDA stream-based IPC communication for multi-GPU systems with and without NVLink
 - Enhanced performance of GPU-based point-to-point communication
 - Leverage Linux Cross Memory Attach (CMA) feature for enhanced host-based communication
 - Enhanced performance of MPI_Allreduce for GPU-resident data
 - InfiniBand Multicast (IB-MCAST) based designs for GPU-based broadcast and streaming applications
 - Basic support for IB-MCAST designs with GPUDirect RDMA
 - Advanced support for zero-copy IB-MCAST designs with GPUDirect RDMA
 - Advanced reliability support for IB-MCAST designs
 - Efficient broadcast designs for Deep Learning applications
 - Enhanced collective tuning on Xeon, OpenPOWER, and NVIDIA DGX-1 systems

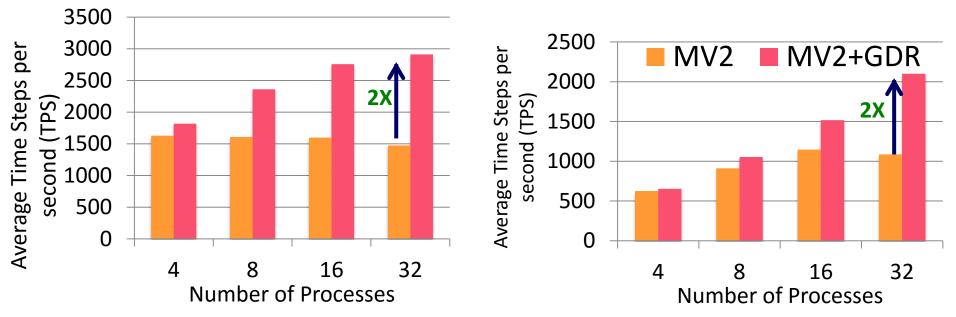
Optimized MVAPICH2-GDR Design



Application-Level Evaluation (HOOMD-blue)

64K Particles

256K Particles

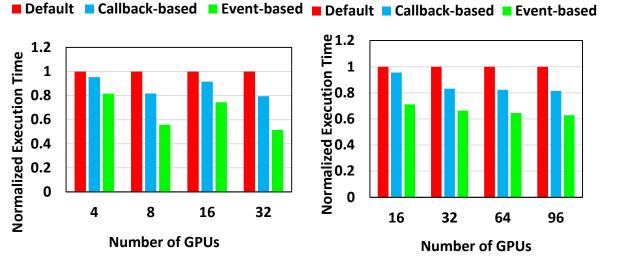


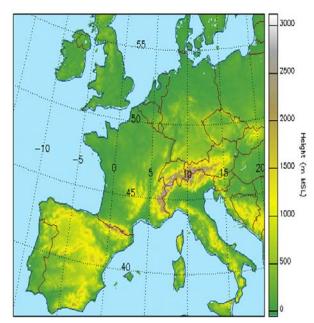
- Platform: Wilkes (Intel Ivy Bridge + NVIDIA Tesla K20c + Mellanox Connect-IB)
- HoomdBlue Version 1.0.5
 - GDRCOPY enabled: MV2_USE_CUDA=1 MV2_IBA_HCA=mlx5_0 MV2_IBA_EAGER_THRESHOLD=32768 MV2_VBUF_TOTAL_SIZE=32768 MV2_USE_GPUDIRECT_LOOPBACK_LIMIT=32768 MV2_USE_GPUDIRECT_GDRCOPY=1 MV2_USE_GPUDIRECT_GDRCOPY_LIMIT=16384

Application-Level Evaluation (Cosmo) and Weather Forecasting in Switzerland



CSCS GPU cluster





- 2X improvement on 32 GPUs nodes
- 30% improvement on 96 GPU nodes (8 GPUs/node)

<u>Cosmo model: http://www2.cosmo-model.org/content</u> /tasks/operational/meteoSwiss/

On-going collaboration with CSCS and MeteoSwiss (Switzerland) in co-designing MV2-GDR and Cosmo Application

C. Chu, K. Hamidouche, A. Venkatesh, D. Banerjee , H. Subramoni, and D. K. Panda, Exploiting Maximal Overlap for Non-Contiguous Data Movement Processing on Modern GPU-enabled Systems, IPDPS'16

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Multi-stream Communication using CUDA IPC on OpenPOWER and DGX-1

Up to 16% higher Device to Device (D2D) bandwidth on OpenPOWER + NVLink inter-connect

Pt-to-pt (D-D) Bandwidth:

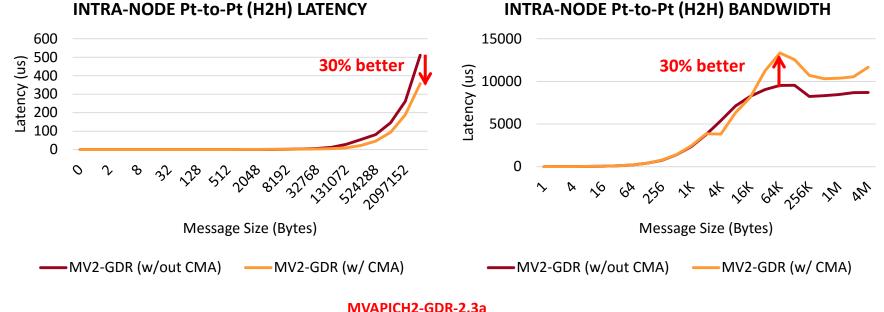
Up to **30% higher** D2D bandwidth on DGX-1 with NVLink

Pt-to-pt (D-D) Bandwidth:

Benefits of Multi-stream CUDA IPC Design Benefits of Multi-stream CUDA IPC Design 20000 40000 Million Bytes (MB)/second Million Bytes (MB)/second 18000 35000 16% better 30% better 16000 30000 14000 25000 12000 10000 20000 8000 15000 6000 10000 4000 5000 2000 0 0 16K 32K 64K 128K 256K 512K 1M 2M 4M 128K 256K 512K 1M 2M 4M Message Size (Bytes) Message Size (Bytes) 1-stream 1-stream 4-streams 4-streams Available with MVAPICH2-GDR-2.3a **Network Based Computing Laboratory** 15

CMA-based Intra-node Communication Support

Up to **30% lower** Host-to-Host (H2H) latency and **30% higher** H2H Bandwidth ۲

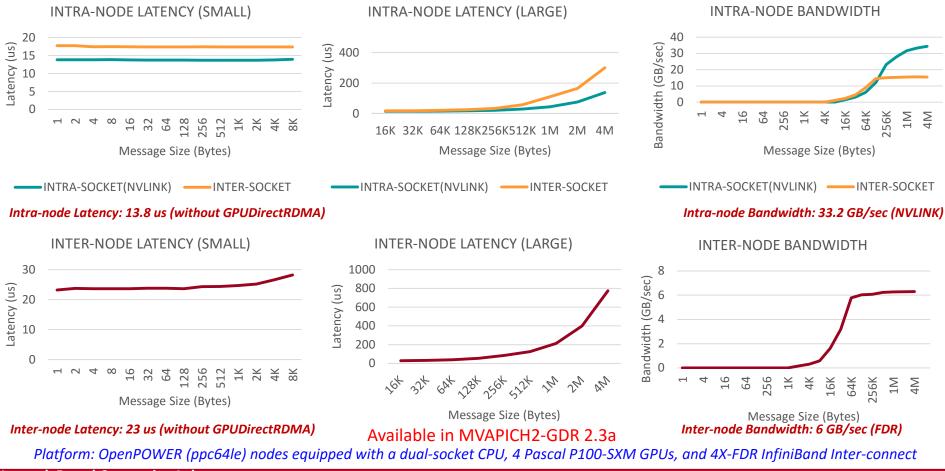


INTRA-NODE Pt-to-Pt (H2H) BANDWIDTH

Intel Broadwell (E5-2680 v4 @ 3240 GHz) node – 28 cores NVIDIA Tesla K-80 GPU, and Mellanox Connect-X4 EDR HCA CUDA 8.0, Mellanox OFED 4.0 with GPU-Direct-RDMA

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MVAPICH2-GDR: Performance on OpenPOWER (NVLink + Pascal)

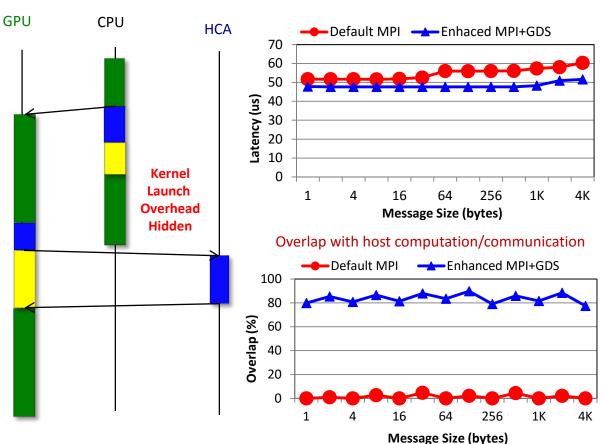


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Control Flow Decoupling through GPUDirect Async

CUDA_Kernel_a<<>>>(A...., stream1) MPI_ISend (A,...., req1, stream1) MPI_Wait (req1, stream1) (non-blocking from CPU) CUDA_Kernel_b<<<>>>(B...., stream1)

- CPU offloads the compute, communication and synchronization tasks to GPU
 - All operations asynchronous from CPU
 - Hide the overhead of kernel launch
- Needs stream-based extensions to MPI semantics
- Latency Oriented: Able to hide the kernel launch overhead - 25% improvement at 256 Bytes
- Throughput Oriented: Asynchronously to offload queue the Communication and computation tasks - 14% improvement at 1KB message size
- Intel Sandy Bridge, NVIDIA K20 and Mellanox FDR HCA
- Will be available in a public release soon



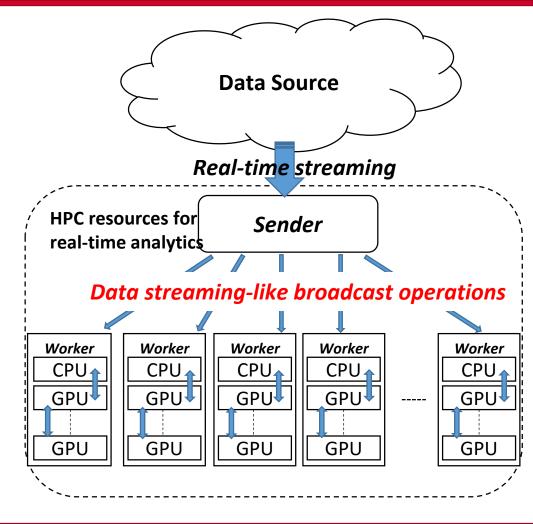
Latency oriented: Kernel+Send and Recv+Kernel

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Streaming Applications

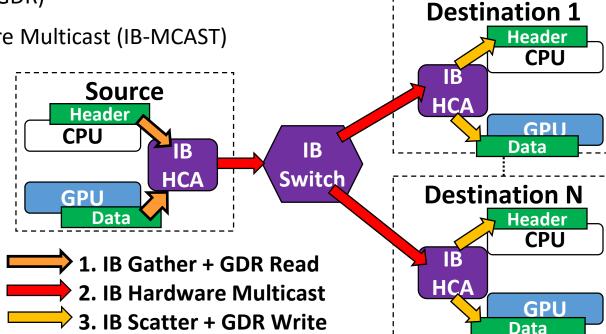
- Streaming applications on HPC systems
 - 1. Communication (MPI)
 - Broadcast-type operations
 - 2. Computation (CUDA)
 - Multiple GPU nodes as workers



Hardware Multicast-based Broadcast

- For GPU-resident data, using \bullet
 - GPUDirect RDMA (GDR)
 - InfiniBand Hardware Multicast (IB-MCAST)
- Overhead
 - **IB UD limit**
 - **GDR** limit

A. Venkatesh, H. Subramoni, K. Hamidouche, and D. K. Panda, "A High Performance Broadcast Design with Hardware Multicast and GPUDirect RDMA for Streaming Applications on InfiniBand Clusters," in HiPC 2014, Dec 2014.



Optimized Broadcast Send

- Preparing Intermediate buffer (*im_buf*)
 - Page-locked (pinned) host buffer

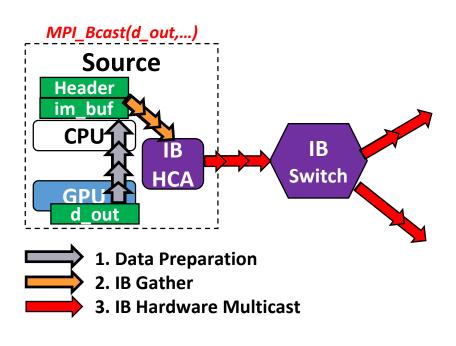
Fast Device-Host data movement

Allocated at initialization phase

Low overhead

- Streaming data through host
 - Fine-tuned chunked data
 - Asynchronous copy operations
 - Three-stage pipeline

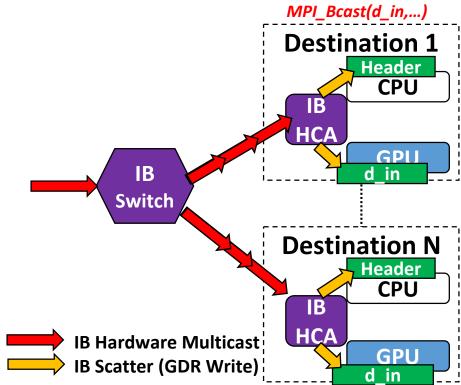
C.-H. Chu, X. Lu, A. A. Awan, H. Subramoni, J. Hashmi, B. Elton and D. K. Panda., "Efficient and Scalable Multi-Source Streaming Broadcast on GPU Clusters for Deep Learning, " ICPP 2017, Aug 14-17, 2017.



Optimized Broadcast Receive

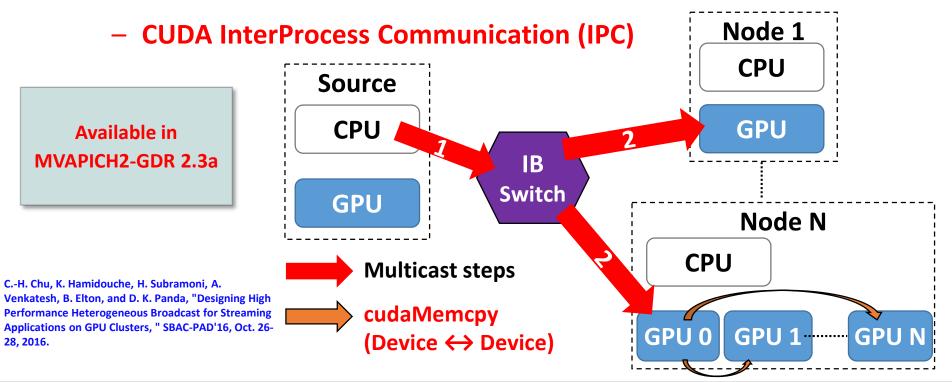
- Zero-copy broadcast receive
 - Pre-posted user buffer (d_in)
 - Avoids additional data movement
 - Leverages IB Scatter and GDR features
 - Low-latency
 - Free-up PCIe resources for applications

C.-H. Chu, X. Lu, A. A. Awan, H. Subramoni, J. Hashmi, B. Elton and D. K. Panda., "Efficient and Scalable Multi-Source Streaming Broadcast on GPU Clusters for Deep Learning, " ICPP 2017, Aug 14-17, 2017.

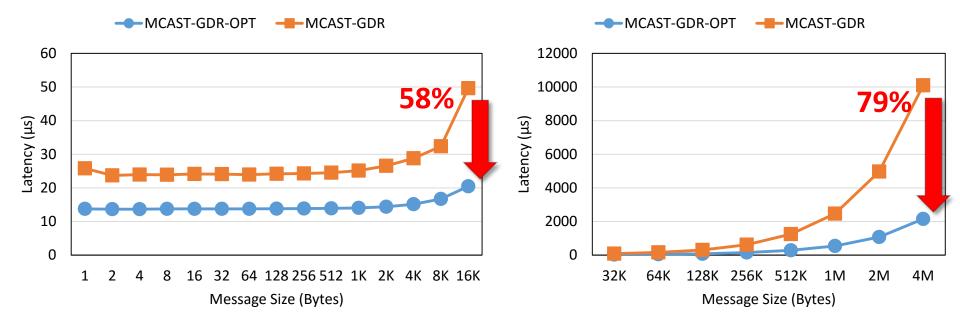


Broadcast on Multi-GPU systems

• Proposed Intra-node Topology-Aware Broadcast



Streaming Benchmark @ CSCS (88 GPUs)



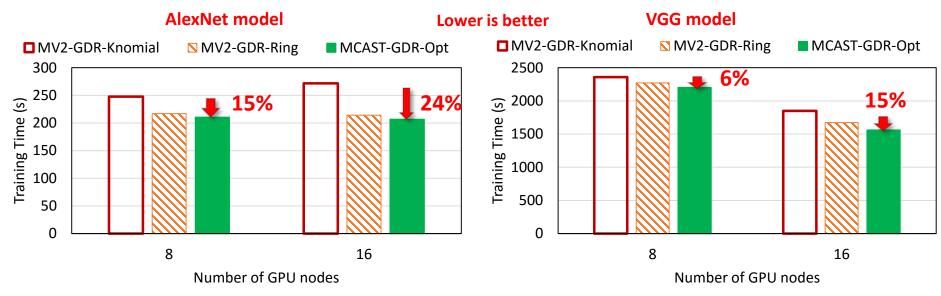
- IB-MCAST + GDR + Topology-aware IPC-based schemes
 - Up to 58% and 79% reduction

for small and large messages

C.-H. Chu, K. Hamidouche, H. Subramoni, A. Venkatesh, B. Elton, and D. K. Panda, "Designing High Performance Heterogeneous Broadcast for Streaming Applications on GPU Clusters, "SBAC-PAD'16, Oct. 26-28, 2016.

Application Evaluation: Deep Learning Frameworks

- @ RI2 cluster, 16 GPUs, 1 GPU/node
 - Microsoft Cognitive Toolkit (CNTK) [https://github.com/Microsoft/CNTK]



- Reduces up to 24% and 15% of latency for AlexNet and VGG models
- Higher improvement can be observed for larger system sizes

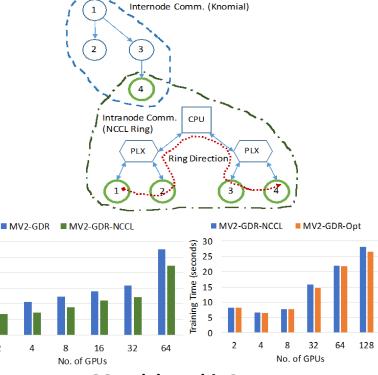
C.-H. Chu, X. Lu, A. A. Awan, H. Subramoni, J. Hashmi, B. Elton, and D. K. Panda, Efficient and Scalable Multi-Source Streaming Broadcast on GPU Clusters for Deep Learning, ICPP'17.

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Efficient Broadcast: MVAPICH2-GDR and NCCL

- NCCL 1.x had some limitations
 - Only worked for a single node; no scale-out on multiple nodes
 - Degradation across IOH (socket) for scale-up (within a node)
- We propose optimized MPI_Bcast design that exploits NCCL^[1]
 - Communication of very large GPU buffers
 - Scale-out on large number of dense multi-GPU nodes
- Hierarchical Communication that efficiently exploits:
 - CUDA-Aware MPI_Bcast in MV2-GDR
 - NCCL Broadcast for intra-node transfers
- Can pure MPI-level designs be done that achieve similar or better performance than NCCL-based approach? ^[2]



VGG Training with CNTK

1. A. A. Awan, K. Hamidouche, A. Venkatesh, and D. K. Panda, Efficient Large Message Broadcast using NCCL and CUDA-Aware MPI for Deep Learning. In *Proceedings of the 23rd European MPI Users' Group Meeting* (EuroMPI 2016). [Best Paper Nominee]

2. A. A. Awan, C-H. Chu, H. Subramoni, and D. K. Panda. Optimized Broadcast for Deep Learning Workloads on Dense-GPU InfiniBand Clusters: MPI or NCCL?, arXiv '17 (<u>https://arxiv.org/abs/1707.09414</u>)

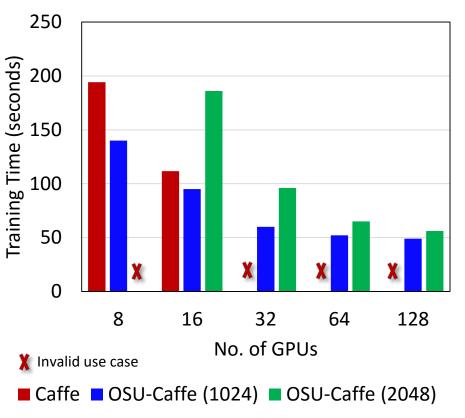
OSU-Caffe 0.9: Scalable Deep Learning on GPU Clusters

SC 2017

- Caffe : A flexible and layered Deep Learning framework.
- Benefits and Weaknesses
 - Multi-GPU Training within a single node
 - Performance degradation for GPUs across different sockets
 - Limited Scale-out
- OSU-Caffe: MPI-based Parallel Training
 - Enable Scale-up (within a node) and Scale-out (across multi-GPU nodes)
 - Scale-out on 64 GPUs for training CIFAR-10 network on CIFAR-10 dataset
 - Scale-out on 128 GPUs for training GoogLeNet network on ImageNet dataset

OSU-Caffe 0.9 available from HiDL site

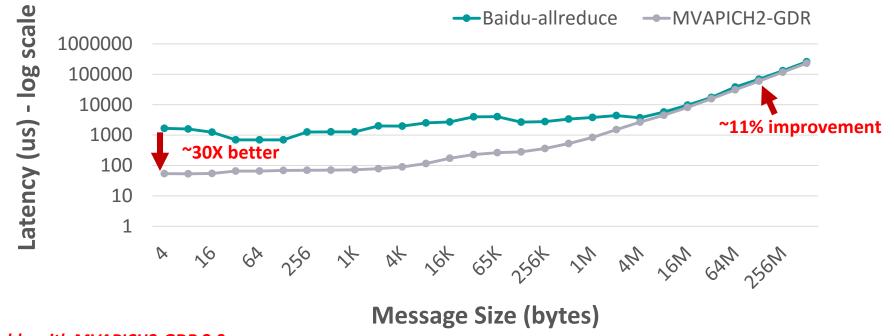
GoogLeNet (ImageNet) on 128 GPUs



Large Message Allreduce: MVAPICH2-GDR vs. Baidu-allreduce

• Performance gains for MVAPICH2-GDR 2.3a* compared to Baidu-allreduce

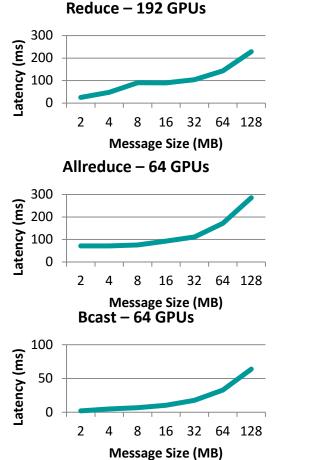
8 GPUs (4 nodes log scale-allreduce vs MVAPICH2-GDR

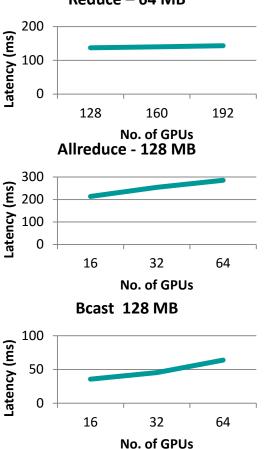


*Available with MVAPICH2-GDR 2.3a

Large Message Optimized Collectives for Deep Learning

- MVAPICH2-GDR provides optimized collectives for large message sizes
- Optimized Reduce, Allreduce, and Bcast
- Good scaling with large number of GPUs
- Available in MVAPICH2-GDR 2.2 and higher





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Conclusions

- MVAPICH2 optimizes MPI communication on InfiniBand clusters with GPUs
- Provides optimized designs for point-to-point two-sided and one-sided communication, datatype processing and collective operations
- Takes advantage of CUDA features like IPC and GPUDirect RDMA families
- Allows flexible solutions for streaming applications with GPUs
- HiDL: Accelerating your Deep Learning framework on HPC systems
 - Tight interaction with MVAPICH2-GDR to boost the performance on GPU cluster
 - Scale-out to multi-GPU nodes

Thank You!

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Network-Based Computing Laboratory http://nowlab.cse.ohio-state.edu/



The High-Performance MPI/PGAS Project <u>http://mvapich.cse.ohio-state.edu/</u>



The High-Performance Deep Learning Project <u>http://hidl.cse.ohio-state.edu/</u>

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