



MVAPICH

MPI, PGAS and Hybrid MPI+PGAS Library

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

Talk at Mellanox Theater (SC '16)

by

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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 2,690 organizations in 83 countries**
 - **More than 402,000 (> 0.4 million) downloads from the OSU site directly**
 - Empowering many TOP500 clusters (Nov '16 ranking)
 - **1st ranked 10,649,640-core cluster (Sunway TaihuLight) at NSC, Wuxi, China**
 - 13th ranked 241,108-core cluster (Pleiades) at NASA
 - 17th ranked 519,640-core cluster (Stampede) at TACC
 - 40th ranked 76,032-core cluster (Tsubame 2.5) at Tokyo Institute of Technology and many others
 - Available with software stacks of many vendors and Linux Distros (RedHat and SuSE)
 - <http://mvapich.cse.ohio-state.edu>
- Empowering Top500 systems for over a decade
 - System-X from Virginia Tech (3rd in Nov 2003, 2,200 processors, 12.25 TFlops) ->
Sunway TaihuLight at NSC, Wuxi, China (1st in Nov'16, 10,649,640 cores, 93 PFlops)



MVAPICH2 Architecture

High Performance Parallel Programming Models

**Message Passing Interface
(MPI)**

**PGAS
(UPC, OpenSHMEM, CAF, UPC++)**

**Hybrid --- MPI + X
(MPI + PGAS + OpenMP/Cilk)**

High Performance and Scalable Communication Runtime

Diverse APIs and Mechanisms

Point-to-point
Primitives

Collectives
Algorithms

Job Startup

Energy-
Awareness

Remote
Memory
Access

I/O and
File Systems

Fault
Tolerance

Virtualization

Active
Messages

Introspection
& Analysis

Support for Modern Networking Technology

(InfiniBand, iWARP, RoCE, OmniPath)

Transport Protocols

RC

XRC

UD

DC

Modern Features

UMR

ODP*

SR-
IOV

Multi
Rail

Support for Modern Multi-/Many-core Architectures

(Intel-Xeon, OpenPower, Xeon-Phi (MIC, KNL*), NVIDIA GPGPU)

Transport Mechanisms

Shared
Memory

CMA

IVSHMEM

Modern Features

MCDRAM*

NVLink*

CAPI*

* Upcoming

MVAPICH2 Software Family

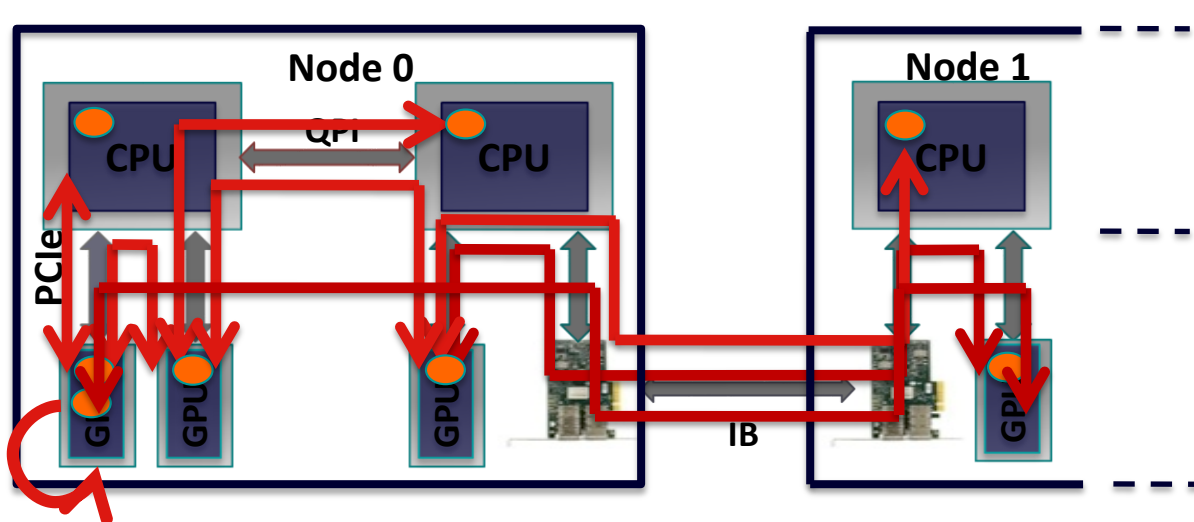
High-Performance Parallel Programming Libraries	
MVAPICH2	Support for InfiniBand, Omni-Path, Ethernet/iWARP, and RoCE
MVAPICH2-X	Advanced MPI features, OSU INAM, PGAS (OpenSHMEM, UPC, UPC++, and CAF), and MPI+PGAS programming models with unified communication runtime
MVAPICH2-GDR	Optimized MPI for clusters 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	
OMB	Microbenchmarks suite to evaluate MPI and PGAS (OpenSHMEM, UPC, and UPC++) libraries for CPUs and GPUs
Tools	
OSU INAM	Network monitoring, profiling, and analysis for clusters with MPI and scheduler integration
OEMT	Utility to measure the energy consumption of MPI applications

Outline

- **MVAPICH2-GPU with GPUDirect-RDMA (GDR)**
- What's new with MVAPICH2-GDR
 - Efficient MPI-3 Non-Blocking Collective support
 - Maximal overlap in MPI Datatype Processing
 - Efficient Support for Managed Memory
 - Initial support for GPUDirect Async feature
- Streaming Support with IB Multicast and GDR
- High-Performance Deep Learning (HiDL) with MVAPICH2-GDR
- Conclusions

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 (\geq CUDA 4.0)
- Overlaps data movement from GPU with RDMA transfers

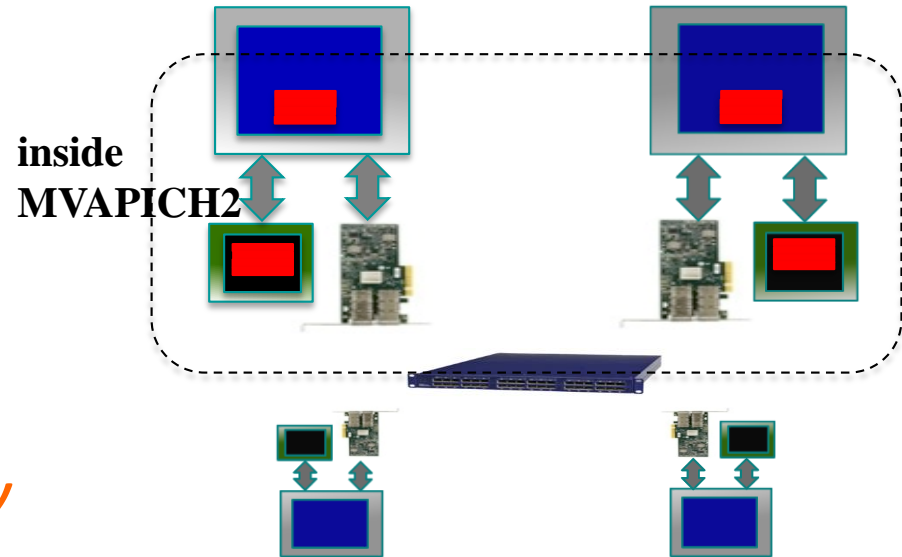
At Sender:

```
MPI_Send(s_devbuf, size, ...);
```

At Receiver:

```
MPI_Recv(r_devbuf, size, ...);
```

High Performance and High Productivity



CUDA-Aware MPI: MVAPICH2-GDR 1.8-2.2 Releases

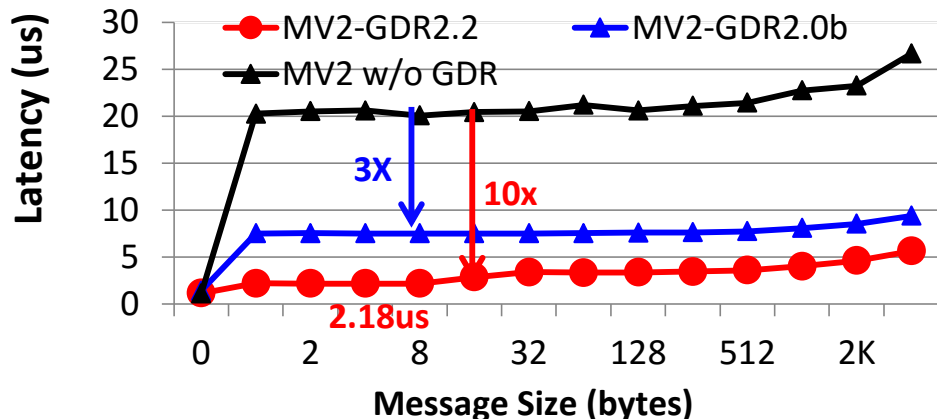
- Support for MPI communication from NVIDIA GPU device memory
- High performance RDMA-based inter-node point-to-point communication (GPU-GPU, GPU-Host and Host-GPU)
- High performance intra-node point-to-point communication for multi-GPU adapters/node (GPU-GPU, GPU-Host and Host-GPU)
- Taking advantage 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

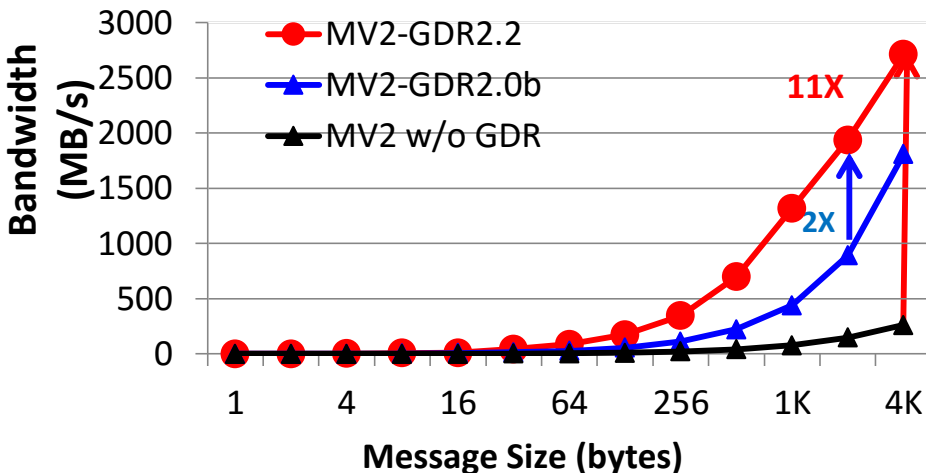
- MVAPICH2-2.2 with GDR support can be downloaded from <https://mvapich.cse.ohio-state.edu/download/mvapich2gdr/>
- System software requirements
 - Mellanox OFED 2.1 or later
 - NVIDIA Driver 331.20 or later
 - NVIDIA CUDA Toolkit 7.0 or later
 - Plugin for GPUDirect RDMA
 - http://www.mellanox.com/page/products_dyn?product_family=116
 - 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

Performance of MVAPICH2-GPU with GPU-Direct RDMA (GDR)

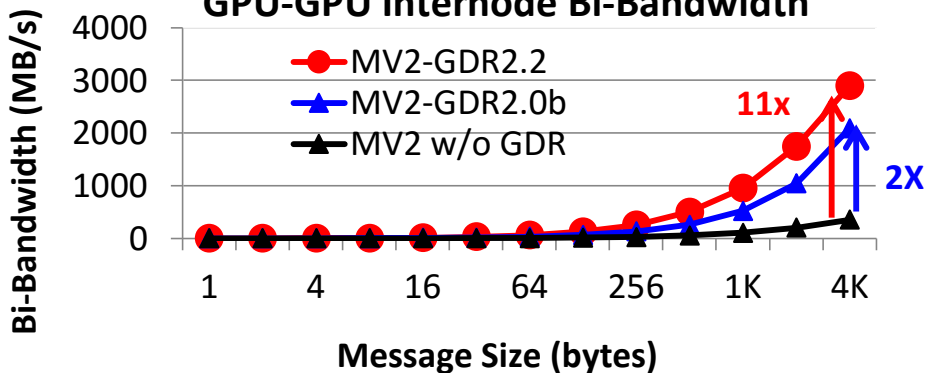
GPU-GPU internode latency



GPU-GPU Internode Bandwidth



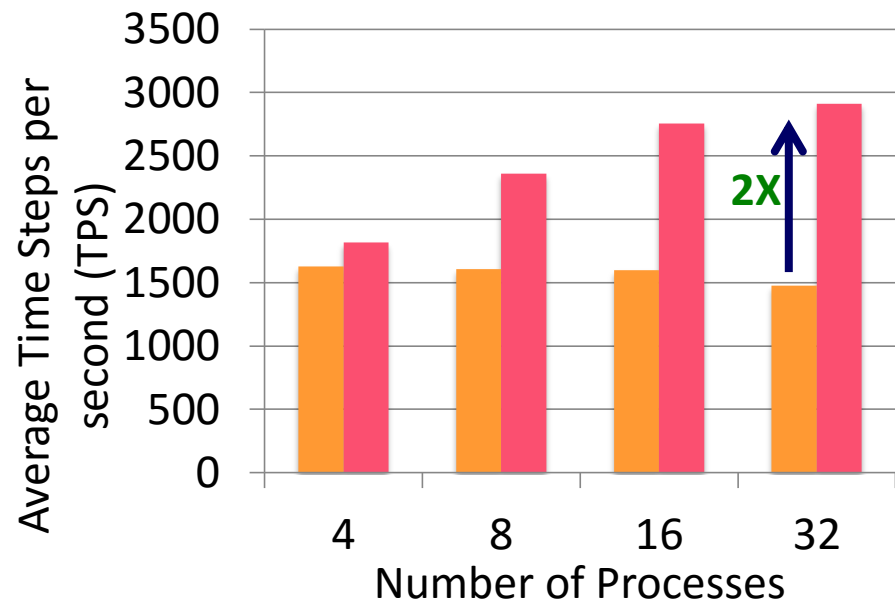
GPU-GPU Internode Bi-Bandwidth



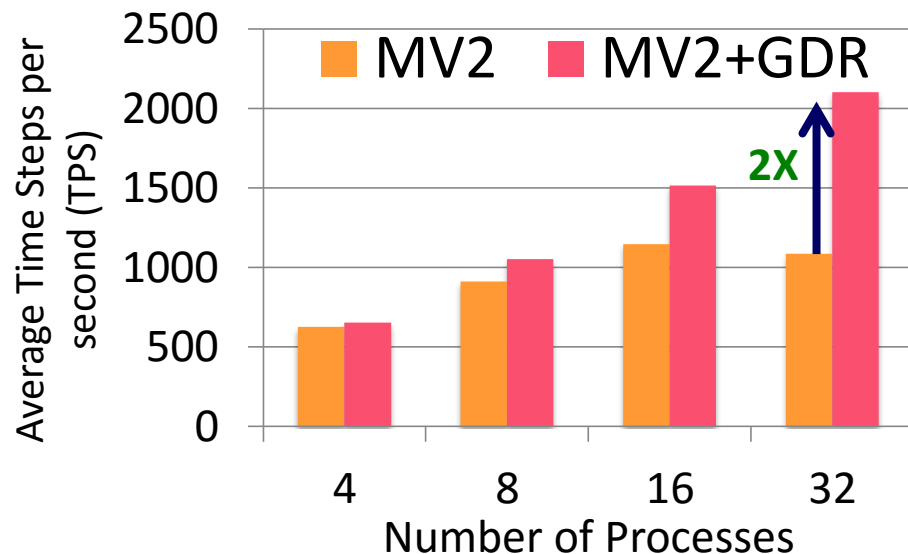
MVAPICH2-GDR-2.2
Intel Ivy Bridge (E5-2680 v2) node - 20 cores
NVIDIA Tesla K40c GPU
Mellanox Connect-X4 EDR HCA
CUDA 8.0
Mellanox OFED 3.0 with GPU-Direct-RDMA

Application-Level Evaluation (HOOMD-blue)

64K Particles

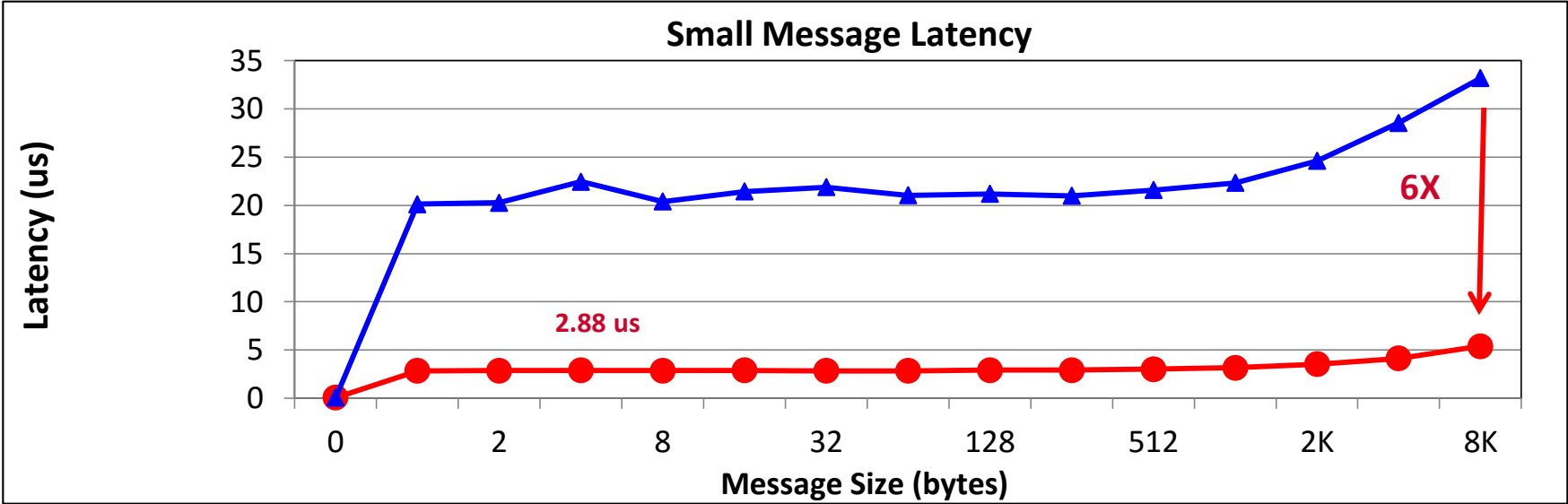


256K Particles



- 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

Full and Efficient MPI-3 RMA Support

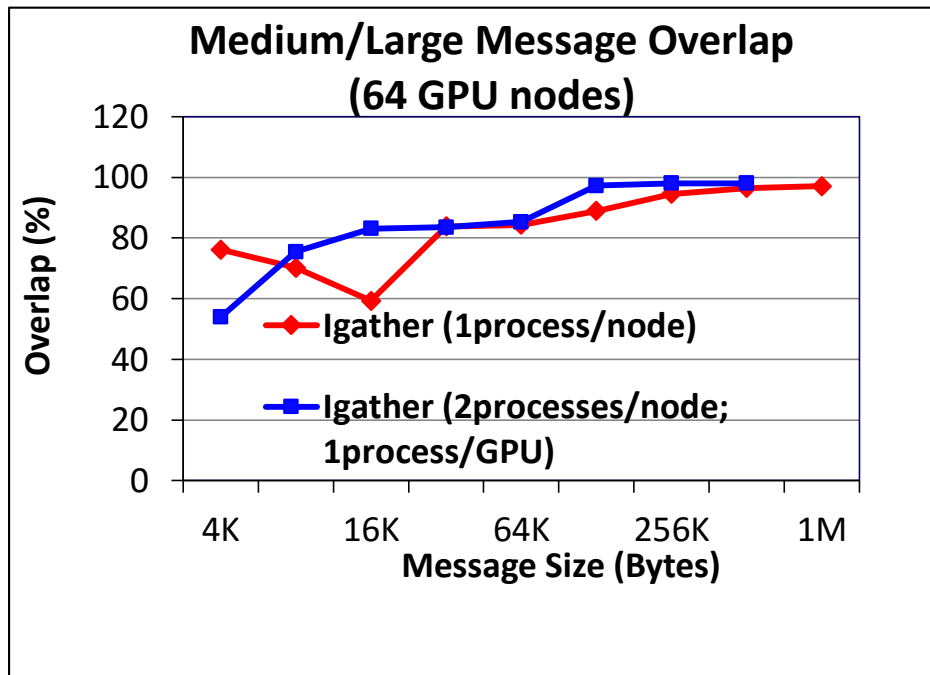
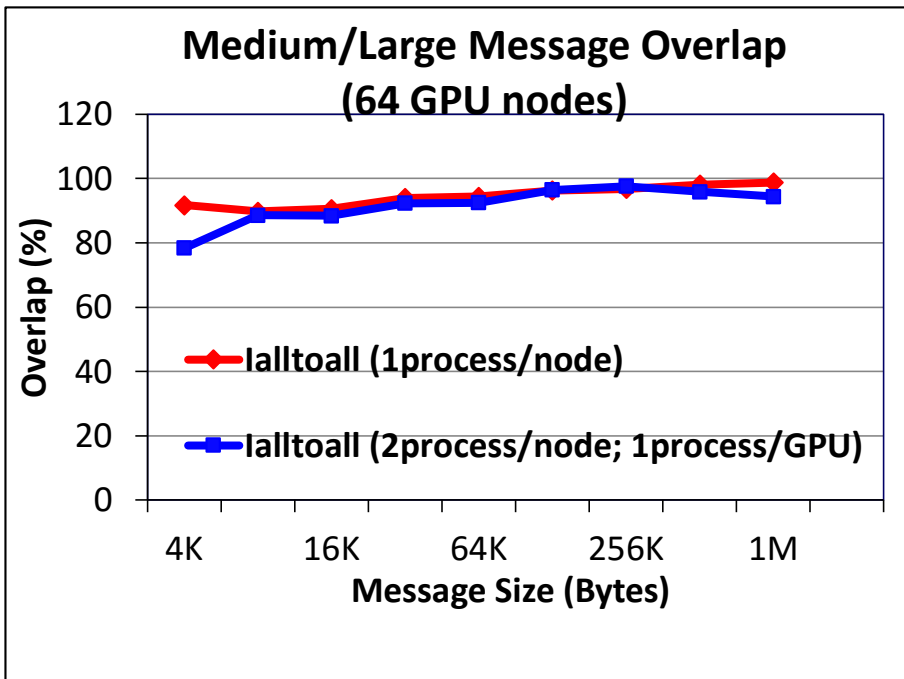


MVAPICH2-GDR-2.2
Intel Ivy Bridge (E5-2680 v2) node - 20 cores, NVIDIA Tesla K40c GPU
Mellanox Connect-IB Dual-FDR HCA, CUDA 7
Mellanox OFED 2.4 with GPU-Direct-RDMA

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CUDA-Aware Non-Blocking Collectives

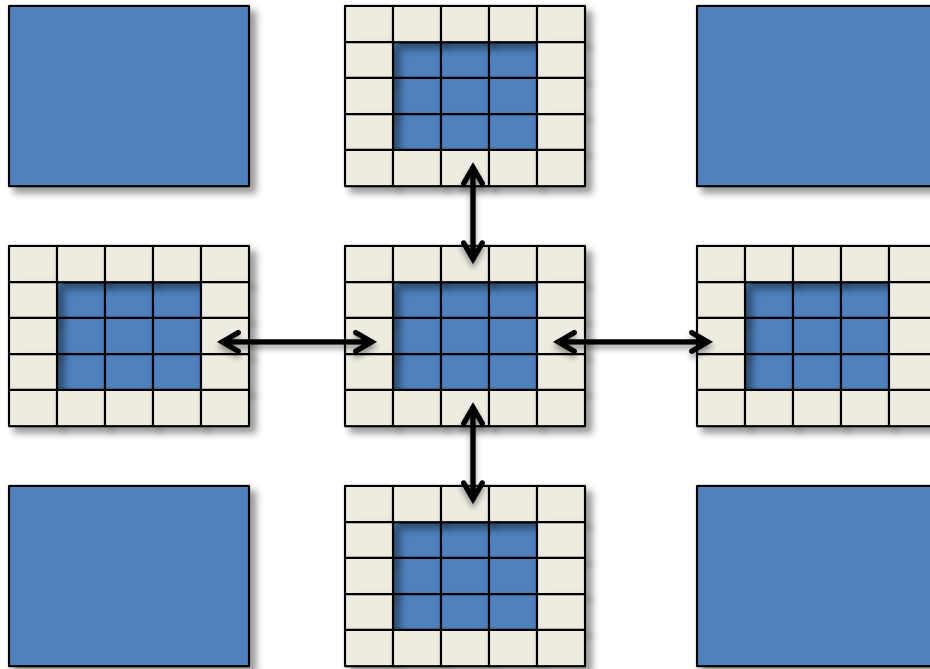


A. Venkatesh, K. Hamidouche, H. Subramoni, and D. K. Panda, Offloaded GPU Collectives using CORE-Direct and CUDA Capabilities on IB Clusters, HIPC, 2015

Platform: Wilkes: Intel Ivy Bridge
NVIDIA Tesla K20c + Mellanox Connect-IB
Available since MVAPICH2-GDR 2.2b

Non-contiguous Data Exchange

Halo data exchange



- Multi-dimensional data
 - Row based organization
 - Contiguous on one dimension
 - Non-contiguous on other dimensions
- Halo data exchange
 - Duplicate the boundary
 - Exchange the boundary in each iteration

MPI Datatype Processing (Computation Optimization)

- Comprehensive support
 - Targeted kernels for regular datatypes - vector, subarray, indexed_block
 - Generic kernels for all other irregular datatypes
- Separate non-blocking stream for kernels launched by MPI library
 - Avoids stream conflicts with application kernels
- Flexible set of parameters for users to tune kernels
 - Vector
 - MV2_CUDA_KERNEL_VECTOR_TIDBLK_SIZE
 - MV2_CUDA_KERNEL_VECTOR_YSIZE
 - Subarray
 - MV2_CUDA_KERNEL_SUBARR_TIDBLK_SIZE
 - MV2_CUDA_KERNEL_SUBARR_XDIM
 - MV2_CUDA_KERNEL_SUBARR_YDIM
 - MV2_CUDA_KERNEL_SUBARR_ZDIM
 - Indexed_block
 - MV2_CUDA_KERNEL_IDXBLK_XDIM

MPI Datatype Processing (Communication Optimization)

Common Scenario

```

MPI_Isend (A,.. Datatype,...)
MPI_Isend (B,.. Datatype,...)
MPI_Isend (C,.. Datatype,...)
MPI_Isend (D,.. Datatype,...)
...

```

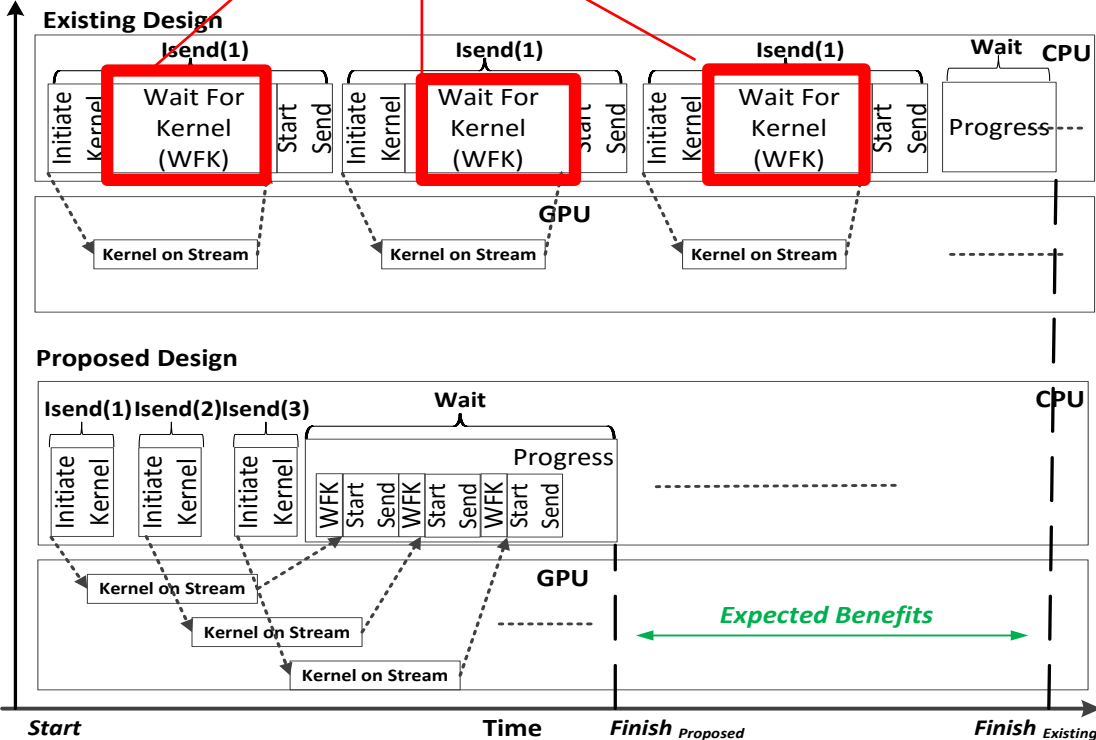
```

MPI_Waitall (...);

```

*A, B...contain non-contiguous MPI Datatype

Waste of computing resources on CPU and GPU



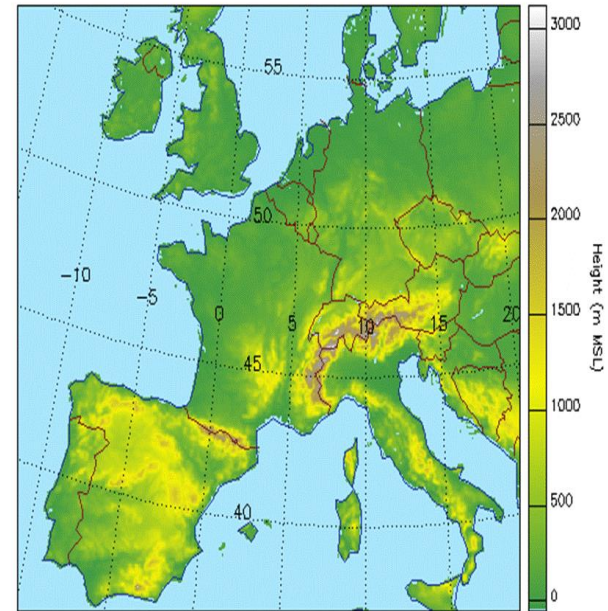
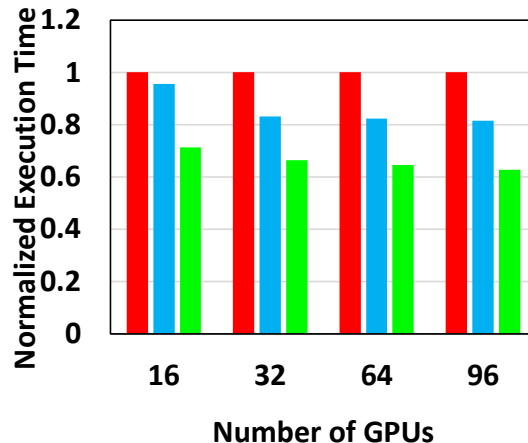
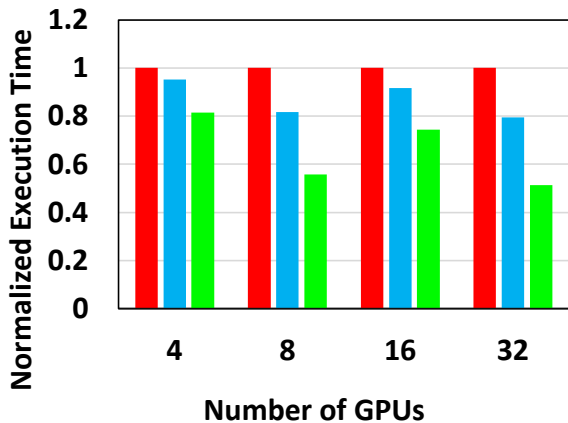
Application-Level Evaluation (Cosmo) and Weather Forecasting in Switzerland

Wilkes GPU Cluster

CSCS GPU cluster

■ Default ■ Callback-based ■ Event-based

■ Default ■ Callback-based ■ Event-based



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

Cosmo model: <http://www2.cosmo-model.org/content/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

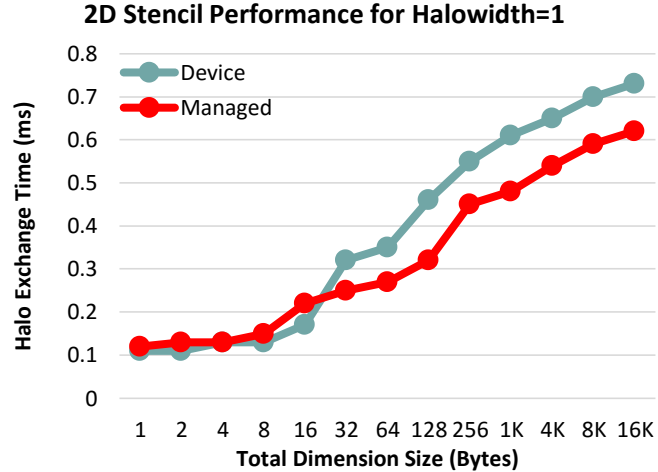
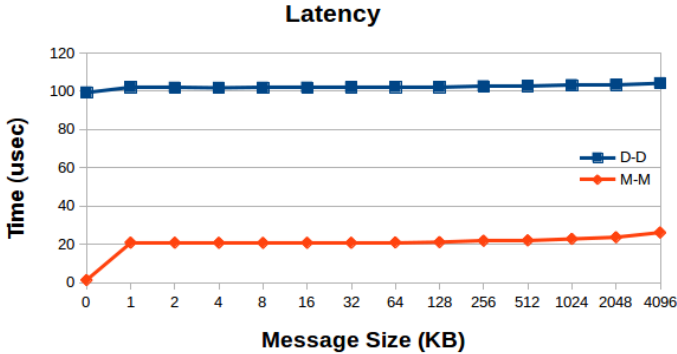
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Initial (Basic) Support for GPU Unified Memory

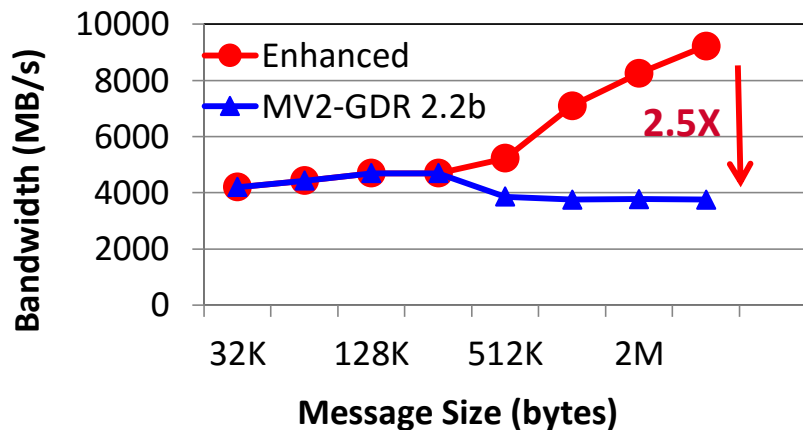
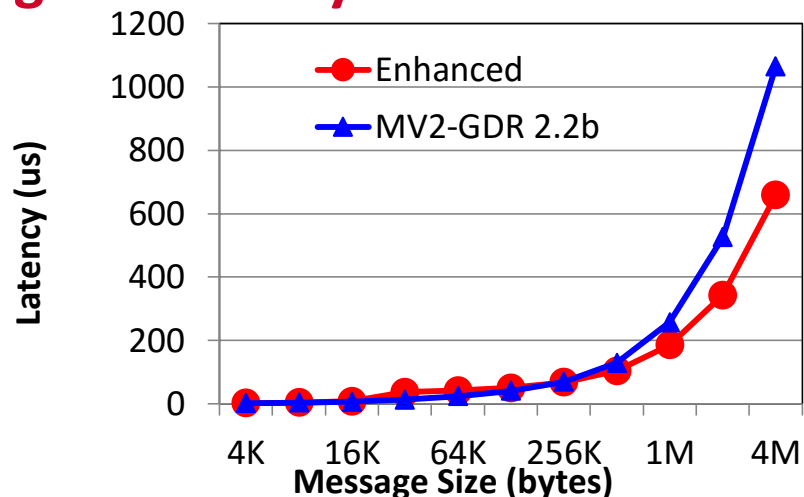
- CUDA 6.0 NVIDIA introduced CUDA Managed (or Unified) memory allowing a common memory allocation for GPU or CPU through *cudaMallocManaged()* call
- Significant productivity benefits due to abstraction of explicit allocation and *cudaMemcpy()*
- Extended MVAPICH2 to perform communications directly from managed buffers (Available since MVAPICH2-GDR 2.2b)
- OSU Micro-benchmarks extended to evaluate the performance of point-to-point and collective communications using managed buffers
 - Available since OMB 5.2

D. S. Banerjee, K Hamidouche, and D. K Panda, Designing High Performance Communication Runtime for GPUManaged Memory: Early Experiences, GPGPU-9 Workshop, held in conjunction with PPOPP '16



Enhanced Support for Intra-node Managed Memory

- CUDA Managed => no memory pin down
 - No IPC support for intra-node communication
 - No GDR support for inter-node communication
- Initial and basic support in MVAPICH2-GDR
 - For both intra- and inter-nodes use “pipeline through” host memory
- Enhance intra-node managed memory to use IPC
 - Double buffering pair-wise IPC-based scheme
 - Brings IPC performance to Managed memory
 - High performance and high productivity
 - 2.5 X improvement in bandwidth
- Available in MVAPICH2-GDR 2.2



Outline

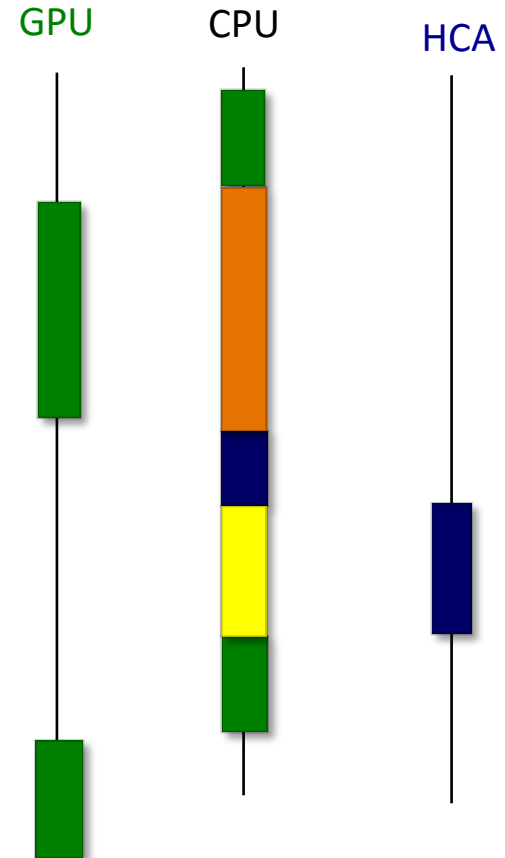
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Overview of GPUDirect aSync (GDS) Feature: Current MPI+CUDA interaction

```
CUDA_Kernel_a<<<>>(A..., stream1)  
cudaStreamSynchronize(stream1)  
MPI_Isend (A..., req1)  
MPI_Wait (req1)  
CUDA_Kernel_b<<<>>(B..., stream1)
```

100% CPU control

- Limits the throughput of a GPU
- Limits the asynchronous progress
- Wastes CPU cycles

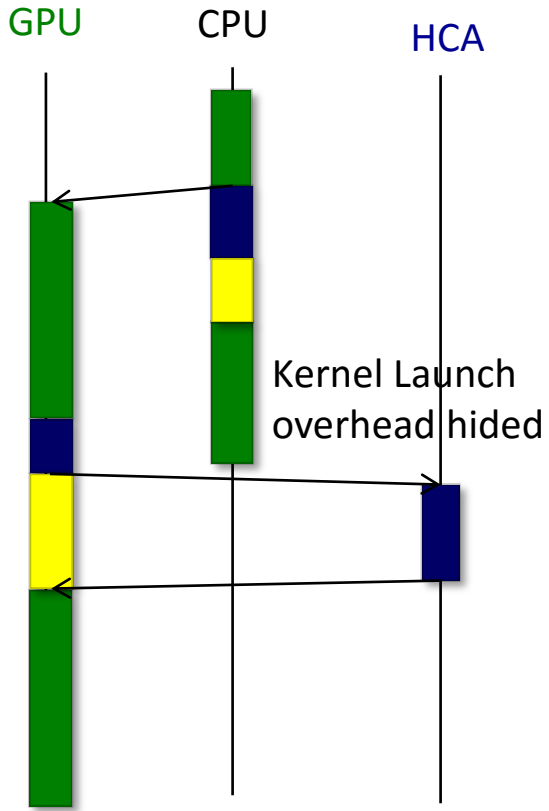


MVAPICH2-GDS: Decouple GPU Control Flow from CPU

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

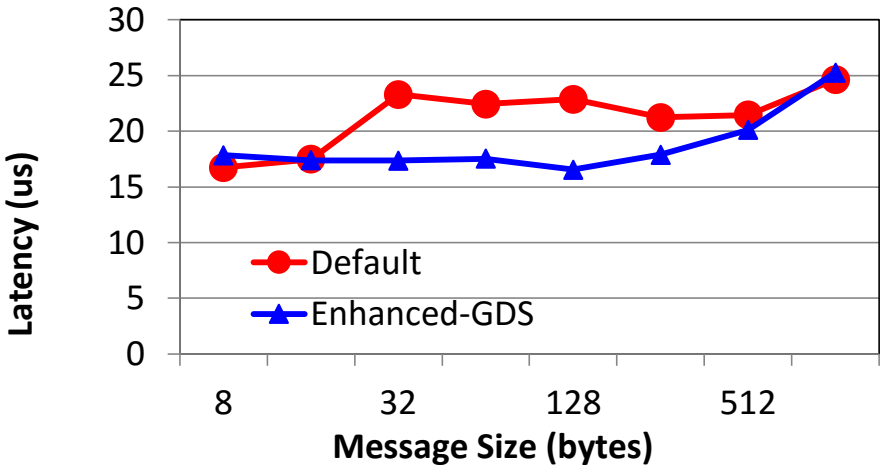
CPU offloads the compute, communication and synchronization tasks to GPU

- CPU is out of the critical path
- Tight interaction between GPU and HCA
- Hides the overhead of kernel launch
- Requires MPI semantics extensions
 - All operations are asynchronous from CPU
 - Extends MPI semantics with Stream-based semantics

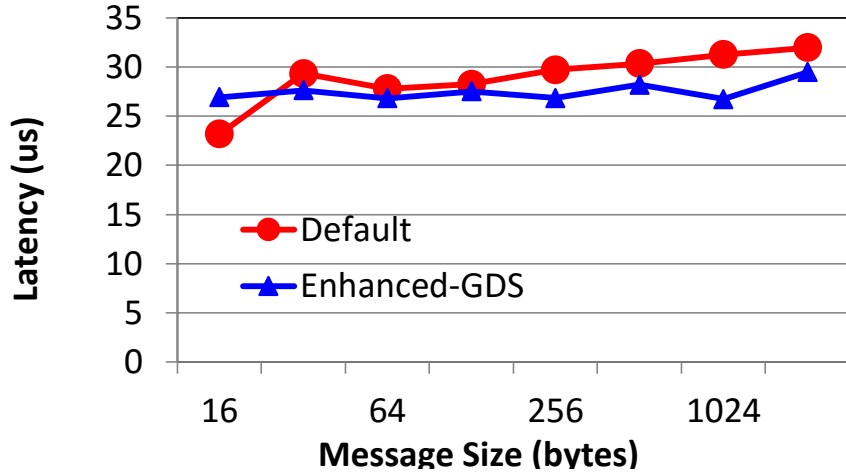


MVAPICH2-GDS: Preliminary Results

Latency oriented: Send+kernel and Recv+kernel



Throughput Oriented: back-to-back



- Latency Oriented: Able to hide the kernel launch overhead
 - 25% improvement at 256 Bytes compared to default behavior
- Throughput Oriented: Asynchronously to offload queue the Communication and computation tasks
 - 14% improvement at 1KB message size
 - Requires some tuning and expect better performance for Application with different Kernels

Intel SandyBridge, NVIDIA K20 and Mellanox FDR HCA

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Streaming Applications

- Examples - surveillance, habitat monitoring, proton computed tomography (pCT), etc..
- Require efficient transport of data from/to distributed sources/sinks
- Sensitive to latency and throughput metrics
- Require HPC resources to efficiently carry out compute-intensive tasks



Src: <http://www.symmetrymagazine.org/article/april-2012/proton-beam-on>

Motivation

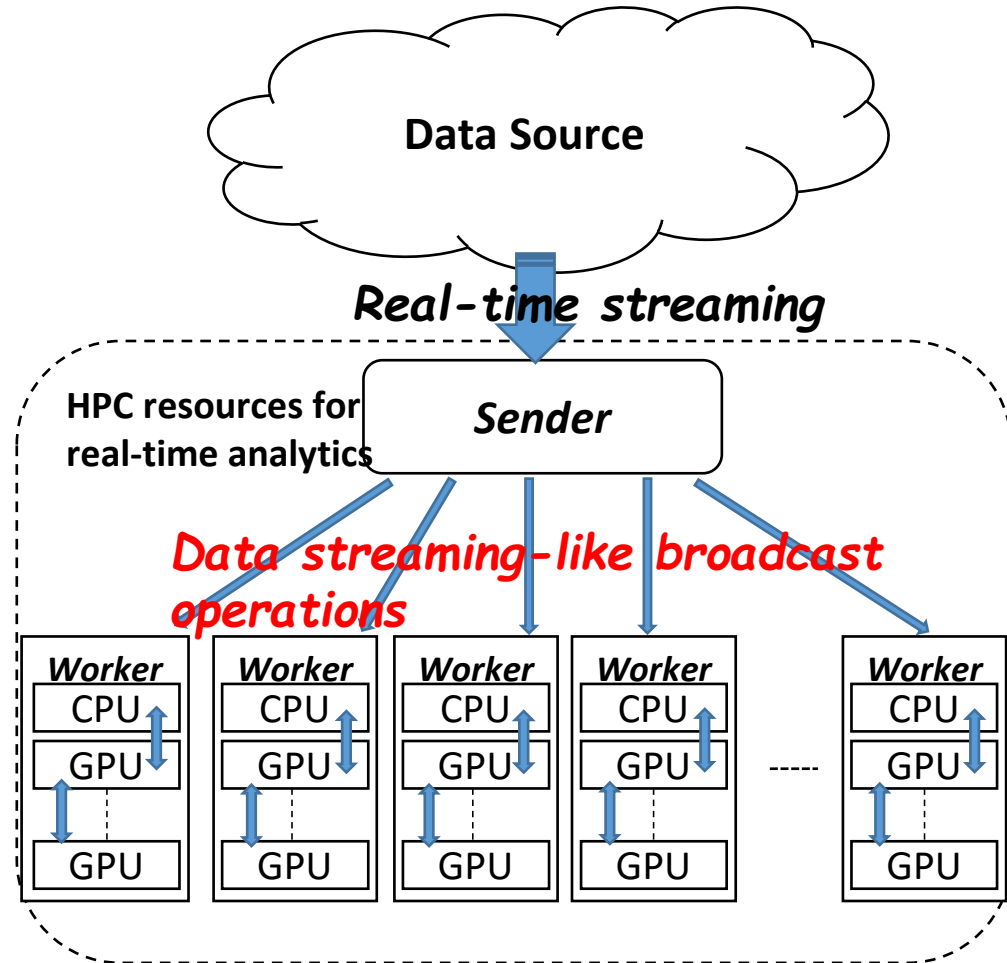
- Streaming applications on HPC systems

1. Communication (**MPI**)

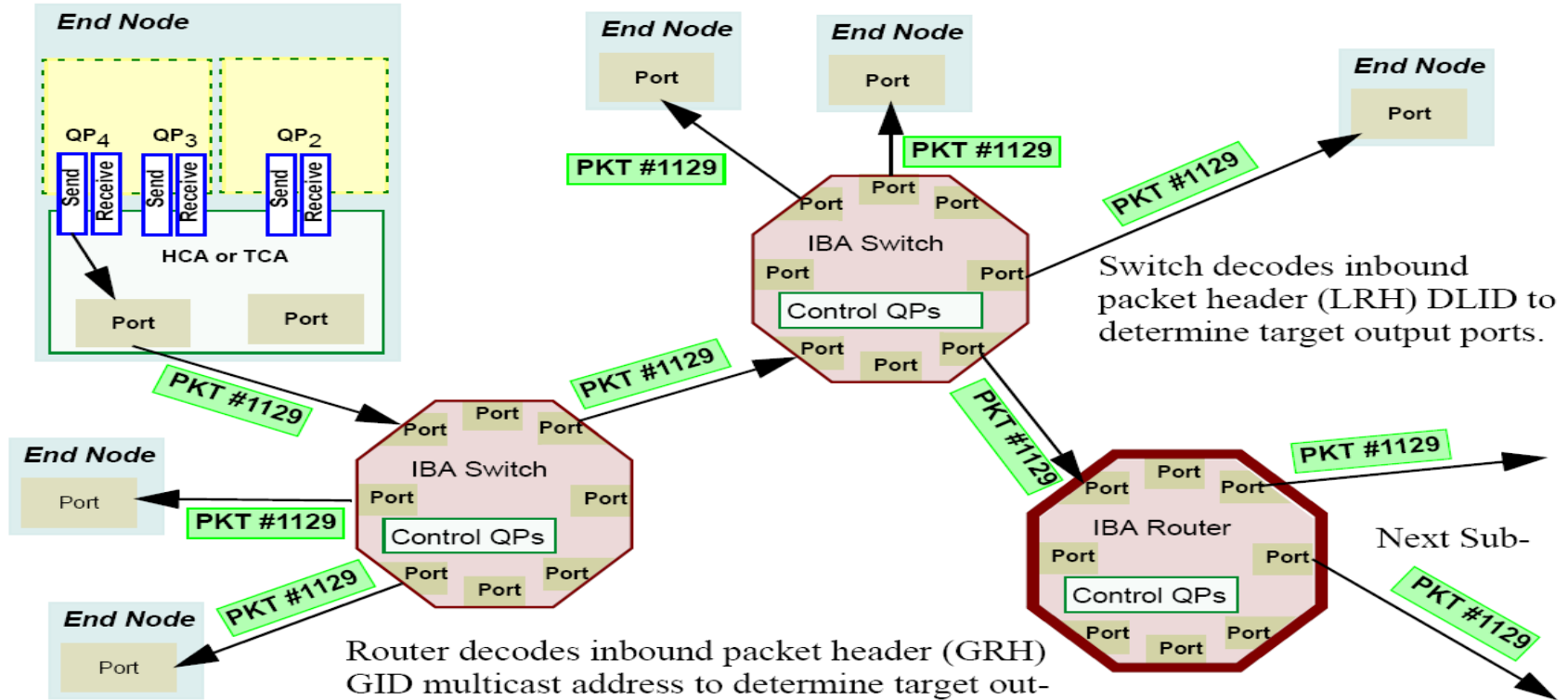
- Broadcast-type operations

2. Computation (**CUDA**)

- Multiple GPU nodes as workers



IB Multicast Example



Problem Statement

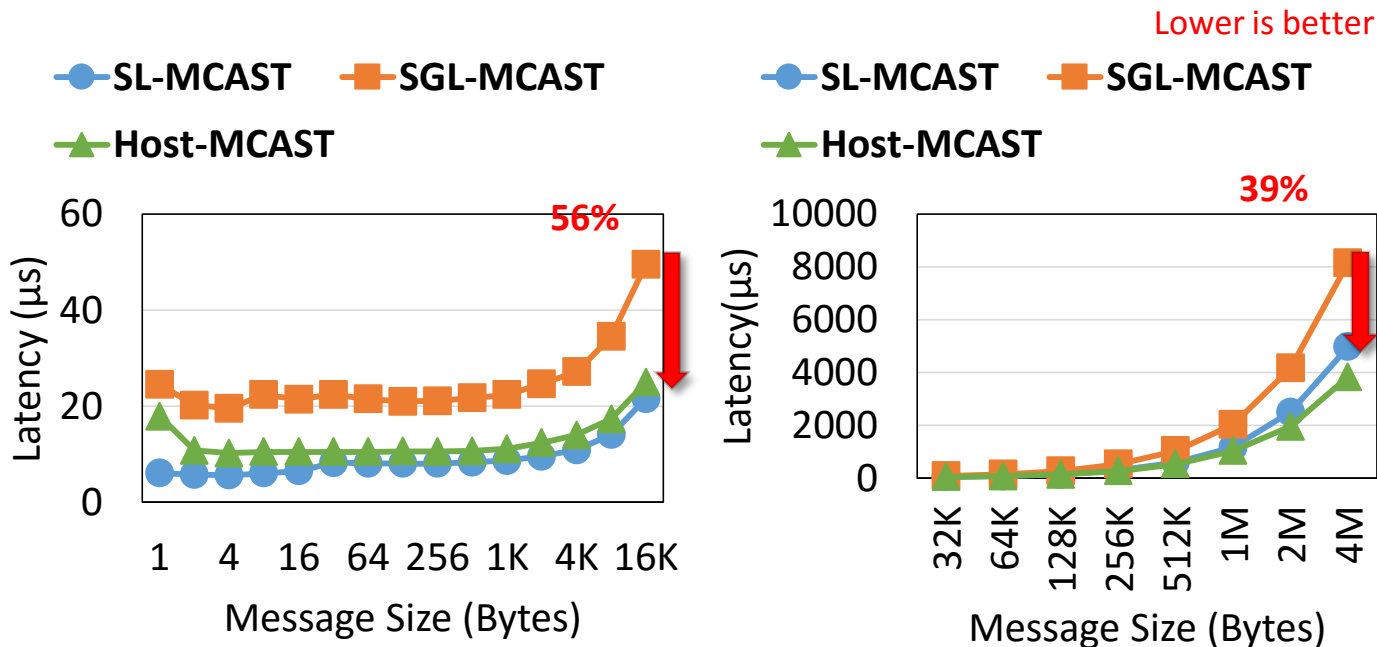
- Can we design a GPU broadcast mechanism that can deliver low latency and high throughput for streaming applications?
- Can we combine GDR and MCAST features to
 - Achieve the best performance
 - Free-up the Host-Device PCIe bandwidth for application needs
- Can such design be extended to support heterogeneous configuration (host-to-device)?
- Can we design an efficient MCAST based broadcast for multi-GPU systems?
- Can we design an efficient reliability support on top of the UD-based MCAST broadcast?
- How can we demonstrate such benefits at benchmark and applications level?

Two Major Solutions (So far)

- **Handling efficient broadcast on multi-GPU node systems**
 - 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 2016.
- **Providing reliability support**
 - C.-H. Chu, K. Hamidouche, H. Subramoni, A. Venkatesh, B. Elton, and D. K. Panda. “Efficient Reliability Support for Hardware Multicast-based Broadcast in GPU-enabled Streaming Applications,“ in COMHPC 2016 (SC Workshop), Nov 2016.
 - Will be presented on Friday (11/18/16) at 9:15am, Room #355-D

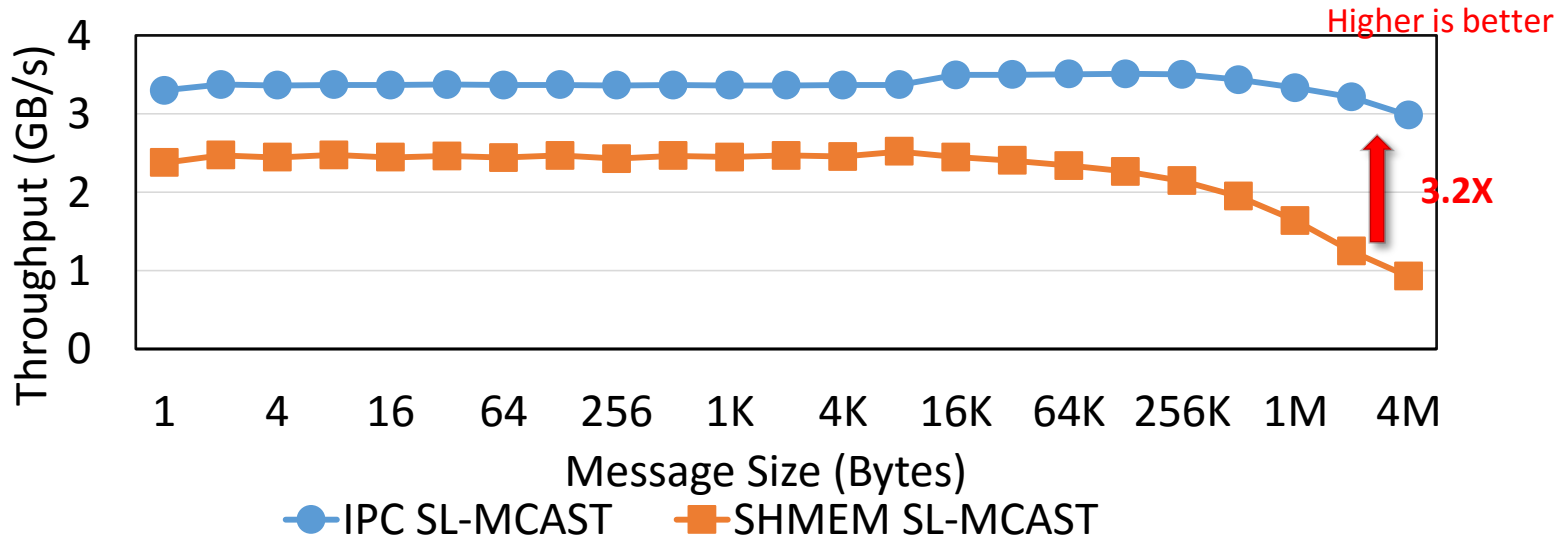
Scatter-List (SL)-based Design for H-D Heterogeneous Support

- Redesigned broadcast benchmark with Root buffer on Host & non-Root on Device
- Inter-node experiments @ Wilkes cluster, 32 GPUs, 1 GPU/node



Benefits of the Availability of Host-Device PCI Resources

- Mimic the behavior of streaming applications @ CSCS cluster, 88 GPUs, 8 NVIDIA K80 GPUs per node
 - Broadcast operations overlapped with application level Host-Device transfers
 - Main thread performing MCAST (streaming)
 - Helper thread starting CUDA kernels and performing Async H-D copies

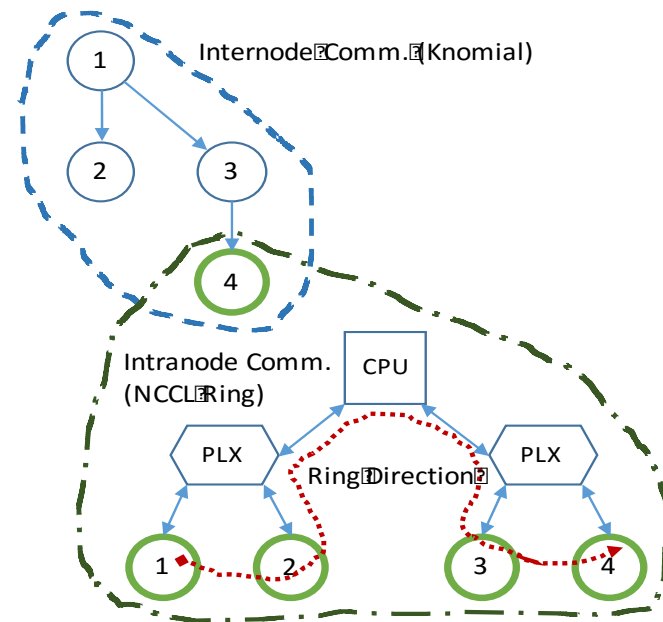


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- **Conclusions**

Deep Learning: New Challenges for MPI Runtimes

- Deep Learning frameworks are a different game altogether
 - Unusually large message sizes (order of megabytes)
 - Most communication based on GPU buffers
- How to address these newer requirements?
 - GPU-specific Communication Libraries (NCCL)
 - Nvidia's NCCL library provides inter-GPU communication
 - CUDA-Aware MPI (MVAPICH2-GDR)
 - Provides support for GPU-based communication
- Can we exploit CUDA-Aware MPI and NCCL to support Deep Learning applications?



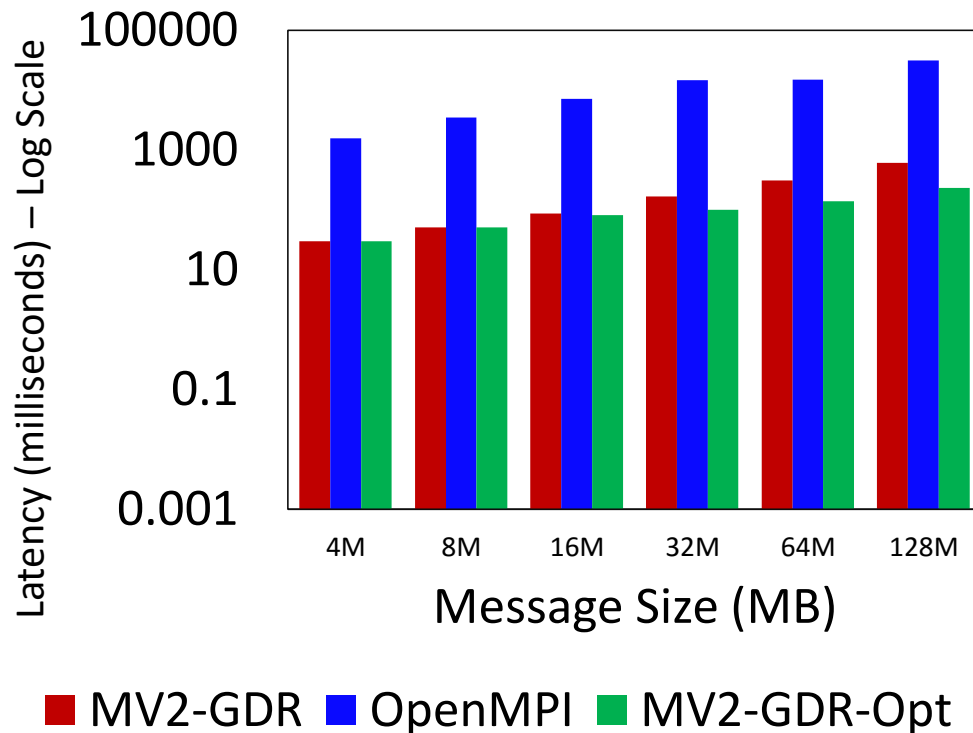
Hierarchical Communication (Knomial + NCCL ring)

Efficient Large Message Broadcast using NCCL and CUDA-Aware MPI for Deep Learning, A. Awan, K. Hamidouche, A. Venkatesh, and D. K. Panda, The 23rd European MPI Users' Group Meeting (EuroMPI 16), Sep 2016 [Best Paper Runner-Up]

Efficient Reduce: MVAPICH2-GDR

- Can we optimize MVAPICH2-GDR to efficiently support DL frameworks?
 - We need to design large-scale reductions using CUDA-Awareness
 - GPU performs reduction using kernels
 - Overlap of computation and communication
 - Hierarchical Designs
- Proposed designs achieve 2.5x speedup over MVAPICH2-GDR and 133x over OpenMPI

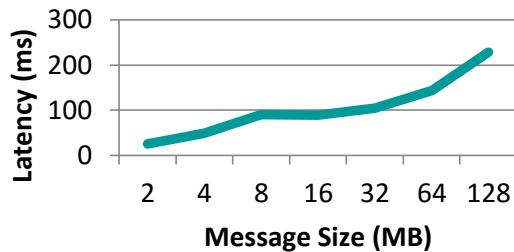
Optimized Large-Size Reduction



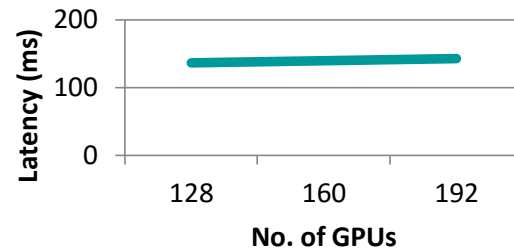
Large Message Optimized Collectives for Deep Learning

- MV2-GDR provides optimized collectives for large message sizes
- Optimized Reduce, Allreduce, and Bcast
- **Good scaling with large number of GPUs**
- **Available in MVAPICH2-GDR 2.2GA**

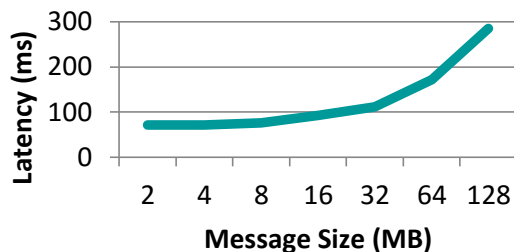
Reduce – 192 GPUs



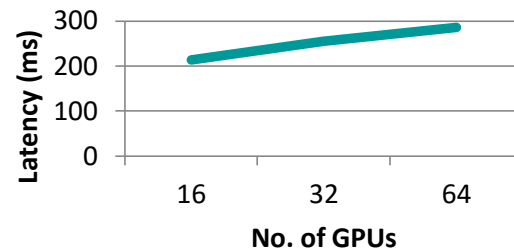
Reduce – 64 MB



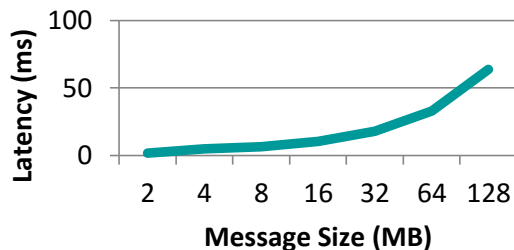
Allreduce – 64 GPUs



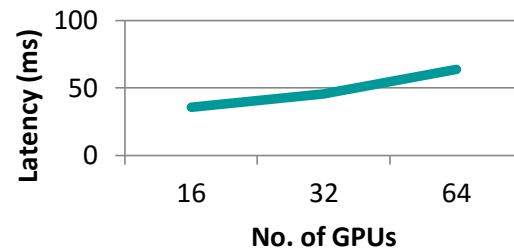
Allreduce - 128 MB



Bcast – 64 GPUs



Bcast 128 MB



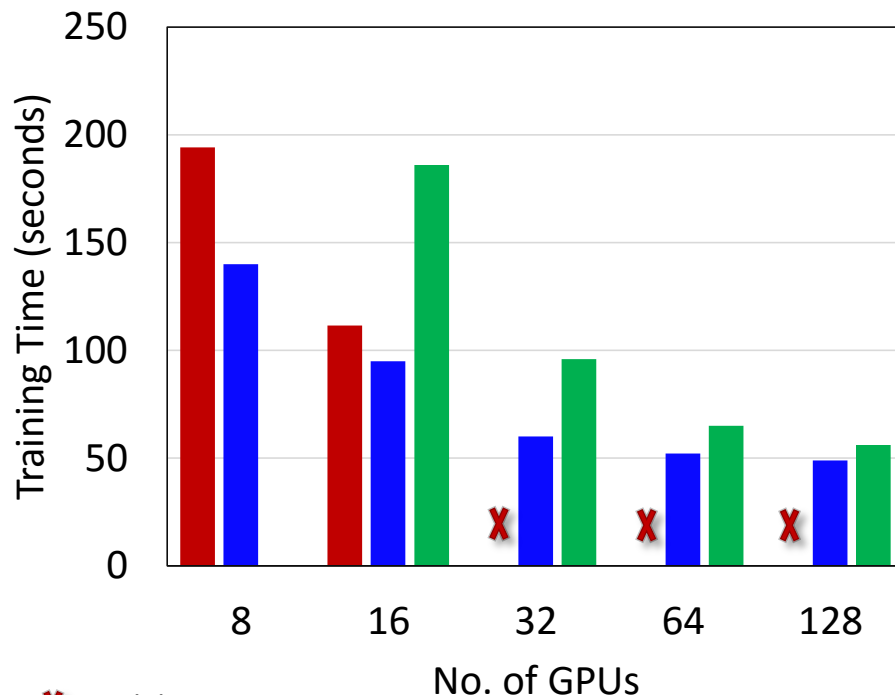
OSU-Caffe: Scalable Deep Learning

- 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 publicly available from

<http://hidl.cse.ohio-state.edu/>

GoogLeNet (ImageNet) on 128 GPUs



X Invalid use case

■ Caffe ■ OSU-Caffe (1024) ■ OSU-Caffe (2048)

Outline

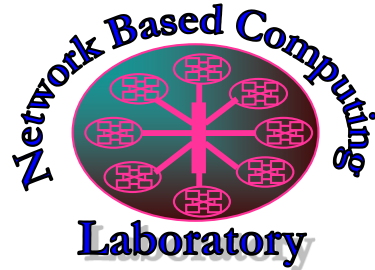
- MVAPICH2-GPU with GPUDirect-RDMA (GDR)
- What's new with MVAPICH2-GDR
 - Efficient MPI-3 Non-Blocking Collective support
 - Maximal overlap in MPI Datatype Processing
 - Efficient Support for Managed Memory
 - Initial support for GPUDirect Async feature
- Streaming Support with IB Multicast and GDR
- High-Performance Deep Learning (HiDL) with MVAPICH2-GDR
- **Conclusions**

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
 - Scales-out to multi-GPU nodes

Thank You!

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Network-Based Computing Laboratory

<http://nowlab.cse.ohio-state.edu/>



The MVAPICH2 Project

<http://mvapich.cse.ohio-state.edu/>