



**MVAPICH**

MPI, PGAS and Hybrid MPI+PGAS Library

# Overview of the MVAPICH Project: Latest Status and Future Roadmap

**MVAPICH2 User Group (MUG) Conference**

**by**

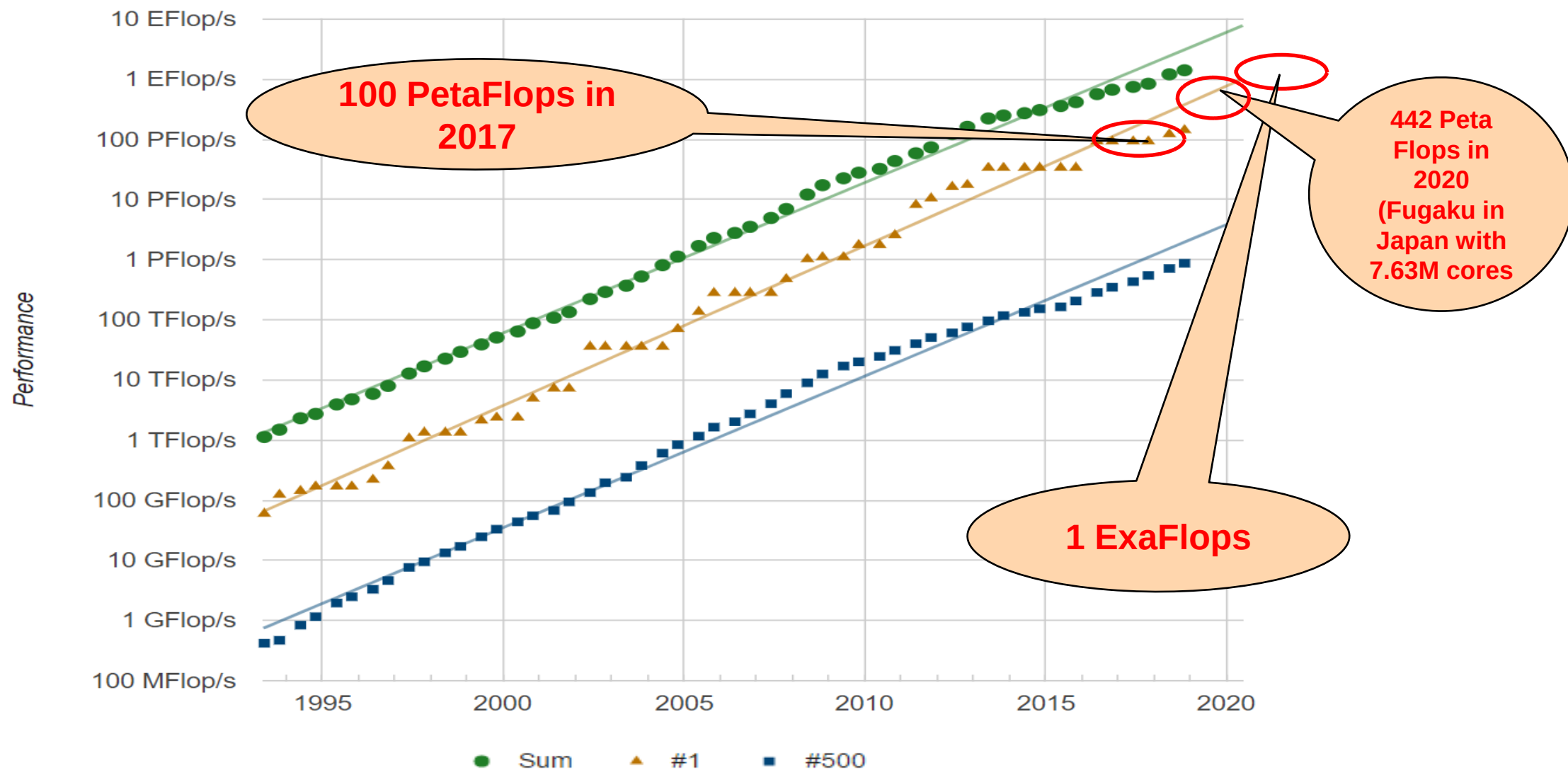
**Dhabaleswar K. (DK) Panda**

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<http://www.cse.ohio-state.edu/~panda>

# High-End Computing (HEC): PetaFlop to ExaFlop



***Expected to have an ExaFlop system in 2021-2022!***

# Supporting Programming Models for Multi-Petaflop and Exaflop Systems: Challenges

**Application Kernels/Applications (HPC and DL)**

**Middleware**

**Programming Models**

MPI, PGAS (UPC, Global Arrays, OpenSHMEM), CUDA, OpenMP, OpenACC, Hadoop, Spark (RDD, DAG), TensorFlow, PyTorch, etc.

**Communication Library or Runtime for Programming Models**

Point-to-point  
Communication

Collective  
Communication

Energy-  
Awareness

Synchronization  
and Locks

I/O and  
File Systems

Fault  
Tolerance

**Networking Technologies**  
(InfiniBand, Ethernet,  
RoCE, Omni-Path, and Slingshot)

**Multi-/Many-core  
Architectures**

**Accelerators  
(GPU and FPGA)**

**Co-Design  
Opportunities  
and Challenges  
across Various  
Layers**

**Performance  
Scalability  
Resilience**

# Designing (MPI+X) at Exascale

- Scalability for million to billion processors
  - Support for highly-efficient inter-node and intra-node communication (both two-sided and one-sided)
  - Scalable job start-up
  - Low memory footprint
- Scalable Collective communication
  - Offload
  - Non-blocking
  - Topology-aware
- Balancing intra-node and inter-node communication for next generation nodes (128-1024 cores)
  - Multiple end-points per node
- Support for efficient multi-threading
- Integrated Support for Accelerators (GPGPUs and FPGAs)
- Fault-tolerance/resiliency
- QoS support for communication and I/O
- Support for Hybrid MPI+PGAS programming (MPI + OpenMP, MPI + UPC, MPI + OpenSHMEM, MPI+UPC++, CAF, ...)
- Virtualization
- Energy-Awareness

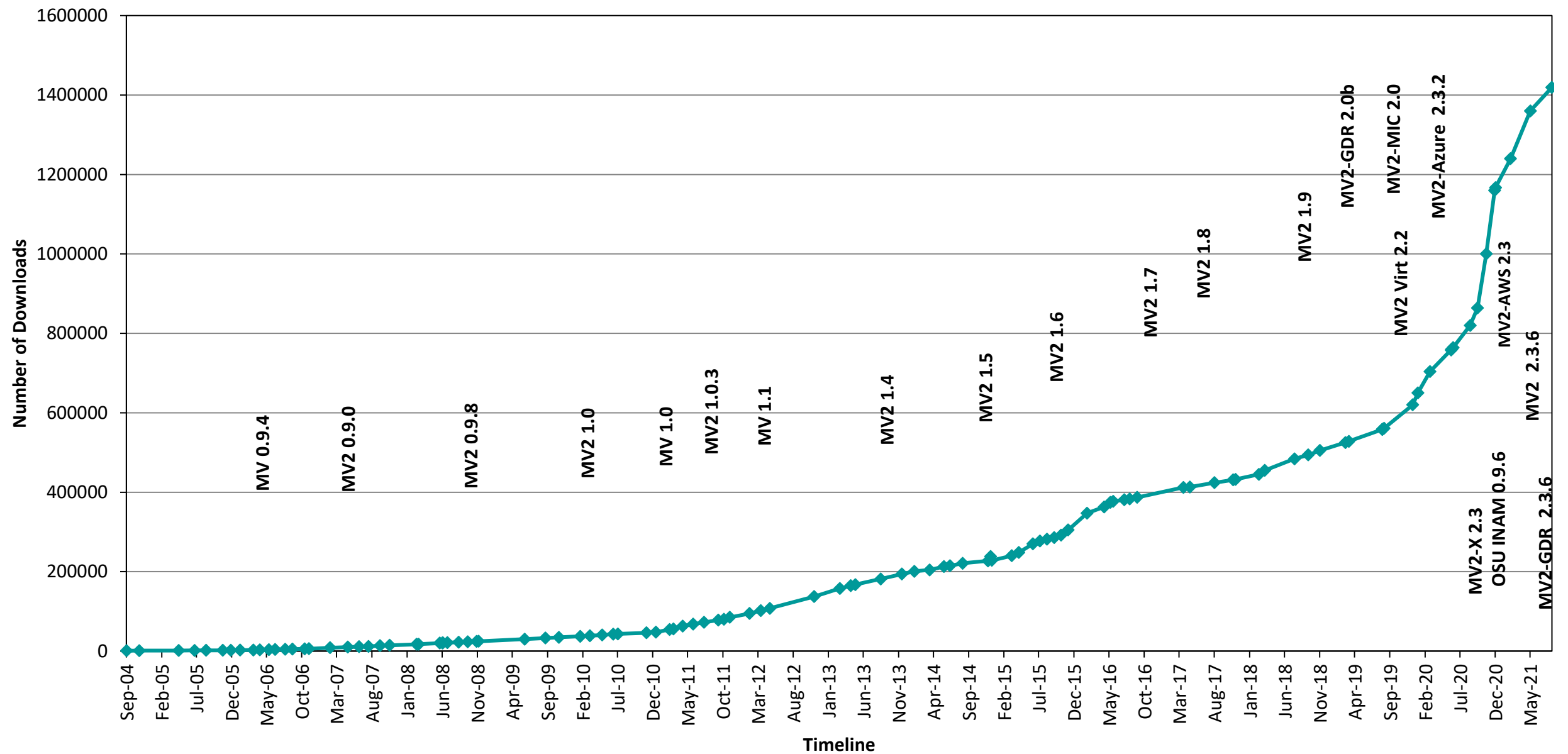
# Overview of the MVAPICH2 Project

- High Performance open-source MPI Library
- Support for multiple interconnects
  - InfiniBand, Omni-Path, Ethernet/iWARP, RDMA over Converged Ethernet (RoCE), and AWS EFA
- Support for multiple platforms
  - x86, OpenPOWER, ARM, Xeon-Phi, GPGPUs (NVIDIA and AMD)
- **Started in 2001, first open-source version demonstrated at SC '02**
- Supports the latest MPI-3.1 standard
- <http://mvapich.cse.ohio-state.edu>
- Additional optimized versions for different systems/environments:
  - MVAPICH2-X (Advanced MPI + PGAS), since 2011
  - MVAPICH2-GDR with support for NVIDIA GPGPUs, since 2014
  - MVAPICH2-MIC with support for Intel Xeon-Phi, since 2014
  - MVAPICH2-Virt with virtualization support, since 2015
  - MVAPICH2-EA with support for Energy-Awareness, since 2015
  - MVAPICH2-Azure for Azure HPC IB instances, since 2019
  - MVAPICH2-X-AWS for AWS HPC+EFA instances, since 2019
- Tools:
  - OSU MPI Micro-Benchmarks (OMB), since 2003
  - OSU InfiniBand Network Analysis and Monitoring (INAM), since 2015

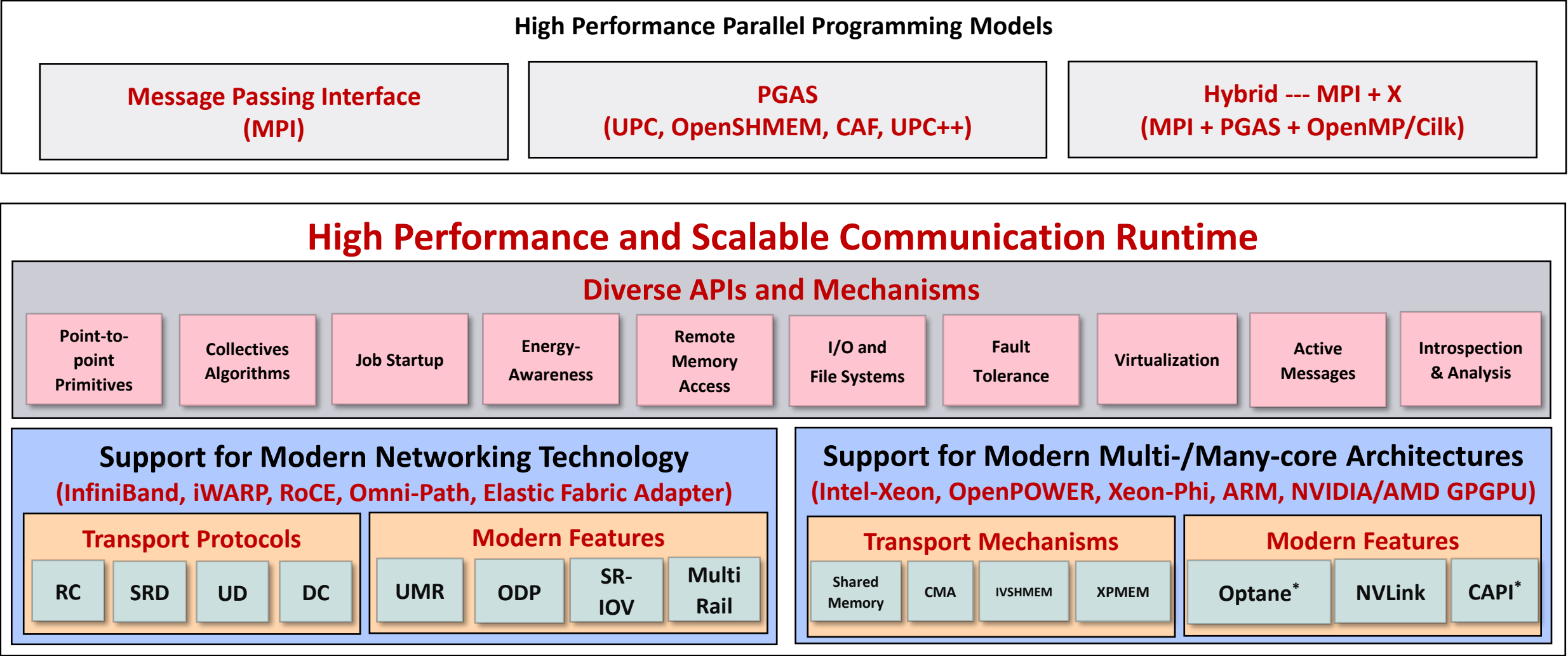


- **Used by more than 3,200 organizations in 89 countries**
- **More than 1.43 Million downloads from the OSU site directly**
- Empowering many TOP500 clusters (June '21 ranking)
  - **4<sup>th</sup> , 10,649,600-core (Sunway TaihuLight) at NSC, Wuxi, China**
  - 10<sup>th</sup>, 448, 448 cores (Frontera) at TACC
  - 20<sup>th</sup>, 288,288 cores (Lassen) at LLNL
  - 31<sup>st</sup>, 570,020 cores (Nurion) in South Korea and many others
- Available with software stacks of many vendors and Linux Distro (RedHat, SuSE, OpenHPC, and Spack)
- Partner in the 10<sup>th</sup> ranked TACC Frontera system
- **Empowering Top500 systems for more than 15 years**

# MVAPICH2 Release Timeline and Downloads



# Architecture of MVAPICH2 Software Family for HPC and DL/ML



\* Upcoming

# Production Quality Software Design, Development and Release

- Rigorous Q&A procedure before making a release
  - Exhaustive unit testing
  - Various test procedures on diverse range of platforms and interconnects
  - Test 19 different benchmarks and applications including, but not limited to
    - OMB, IMB, MPICH Test Suite, Intel Test Suite, NAS, Scalapak, and SPEC
  - Spend about 18,000 core hours per commit
  - Performance regression and tuning
  - Applications-based evaluation
  - Evaluation on large-scale systems
- All versions (alpha, beta, RC1 and RC2) go through the above testing



# MVAPICH2 Software Family

| Requirements                                                                                                                           | Library               |
|----------------------------------------------------------------------------------------------------------------------------------------|-----------------------|
| <b>MPI with Support for InfiniBand, Omni-Path, Ethernet/iWARP and, RoCE (v1/v2)</b>                                                    | <b>MVAPICH2</b>       |
| <b>Optimized Support for Microsoft Azure Platform with InfiniBand</b>                                                                  | <b>MVAPICH2-Azure</b> |
| <b>Advanced MPI features/support (UMR, ODP, DC, Core-Direct, SHArP, XPMEM), OSU INAM (InfiniBand Network Monitoring and Analysis),</b> | <b>MVAPICH2-X</b>     |
| <b>Advanced MPI features (SRD and XPMEM) with support for Amazon Elastic Fabric Adapter (EFA)</b>                                      | <b>MVAPICH2-X-AWS</b> |
| <b>Optimized MPI for clusters with NVIDIA GPUs and for GPU-enabled Deep Learning Applications</b>                                      | <b>MVAPICH2-GDR</b>   |
| <b>Energy-aware MPI with Support for InfiniBand, Omni-Path, Ethernet/iWARP and, RoCE (v1/v2)</b>                                       | <b>MVAPICH2-EA</b>    |
| <b>MPI Energy Monitoring Tool</b>                                                                                                      | <b>OEMT</b>           |
| <b>InfiniBand Network Analysis and Monitoring</b>                                                                                      | <b>OSU INAM</b>       |
| <b>Microbenchmarks for Measuring MPI and PGAS Performance</b>                                                                          | <b>OMB</b>            |

# MVAPICH2 2.3.6

- Released on 05/11/2021
- Major Features and Enhancements
  - Support collective offload using Mellanox's SHARP for Reduce and Bcast
    - Enhanced tuning framework for Reduce and Bcast using SHARP
  - Enhanced performance for UD-Hybrid code
  - Add multi-rail support for UD-Hybrid code
  - Enhanced performance for shared-memory collectives
  - Enhanced job-startup performance for flux job launcher
  - Add support in mpirun\_rsh to use srun daemons to launch jobs
  - Add support in mpirun\_rsh to specify processes per node using '-ppn' option
  - Use PMI2 by default when SLURM is selected as process manager
  - Add support to use aligned memory allocations for multi-threaded applications
  - Architecture detection and enhanced point-to-point tuning for Oracle BM.HPC2 cloud shape
  - Enhanced collective tuning for Frontera@TACC and Expanse@SDSC
  - Add support for GCC compiler v11
  - Add support for Intel IFX compiler
  - Update hwloc v1 code to v1.11.14 & hwloc v2 code to v2.4.2

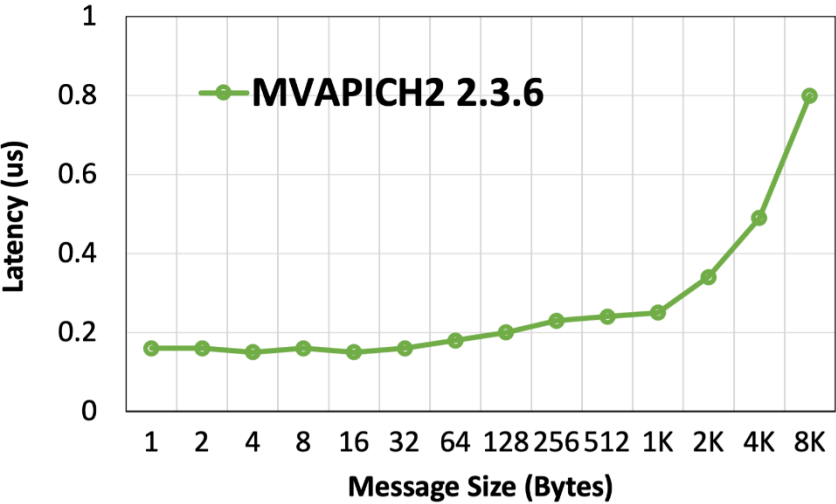
# Highlights of MVAPICH2 2.3.6-GA Release

- Support for highly-efficient inter-node and intra-node communication
- Scalable Start-up
- Collective offload using Mellanox's SHARP support
- Performance Engineering with MPI\_T

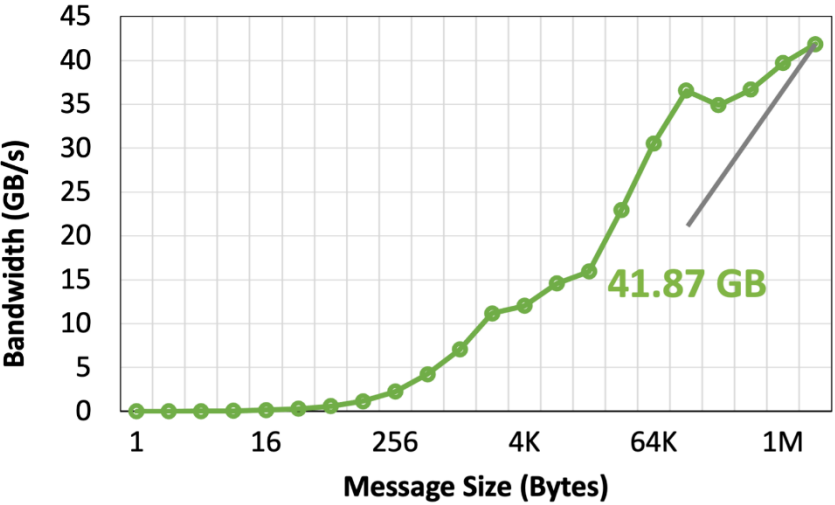
# AMD Milan + HDR 200

## Intra-Node CPU Point-to-Point

Latency

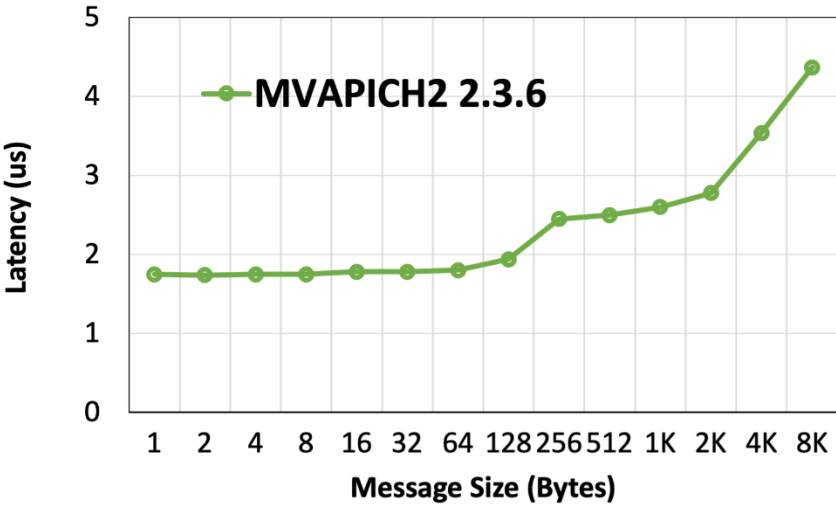


Bandwidth

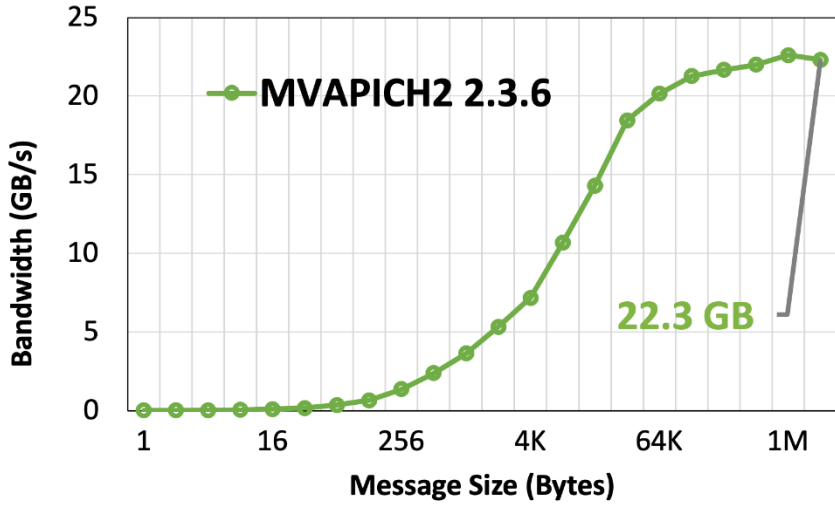


## Inter-Node CPU Point-to-Point

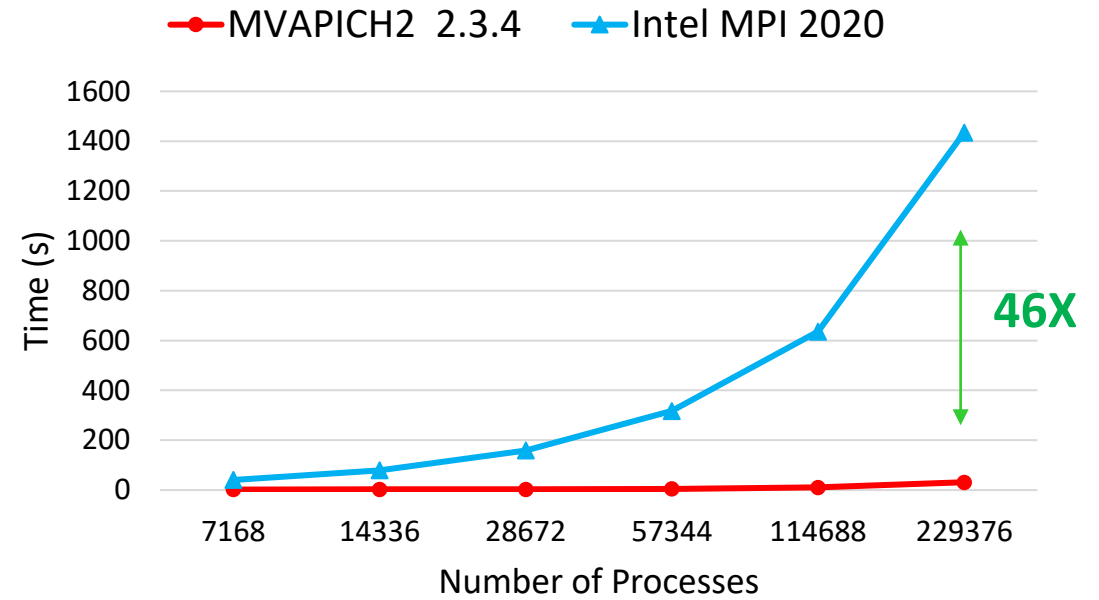
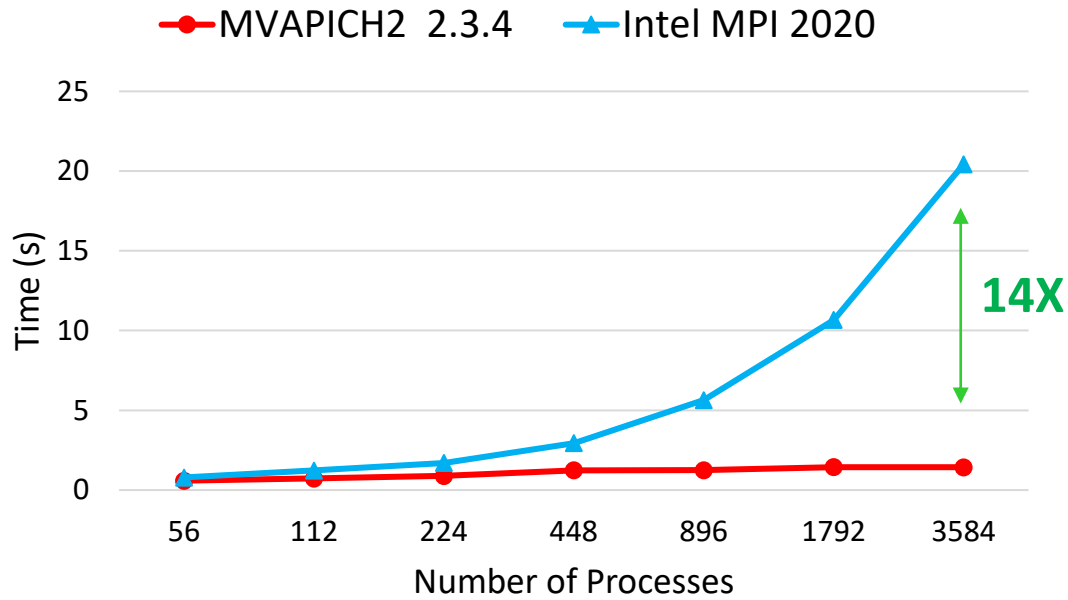
Latency



Bandwidth



# Startup Performance on TACC Frontera

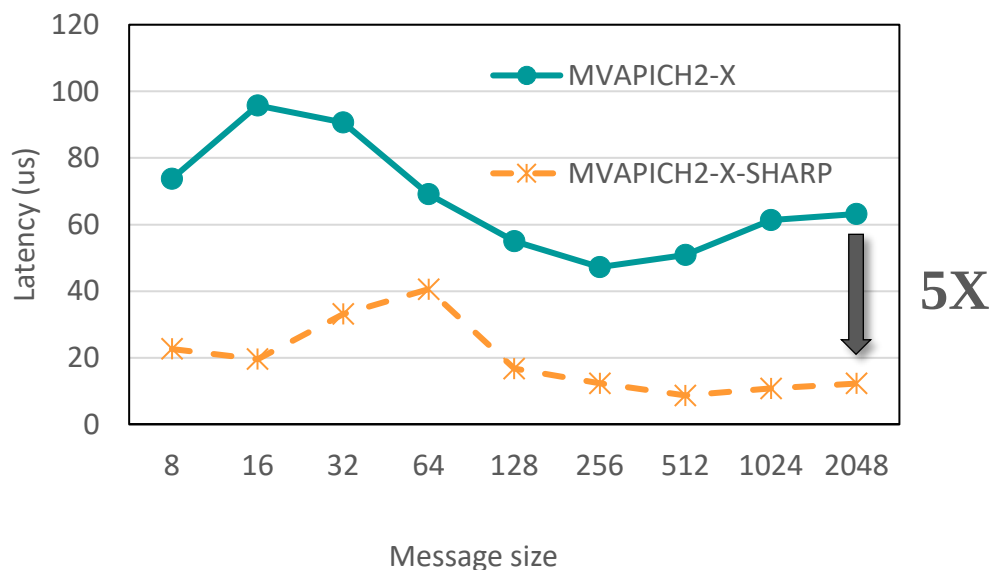


- MPI\_Init takes 31 seconds on 229,376 processes on 4,096 nodes
- All numbers reported with 56 processes per node

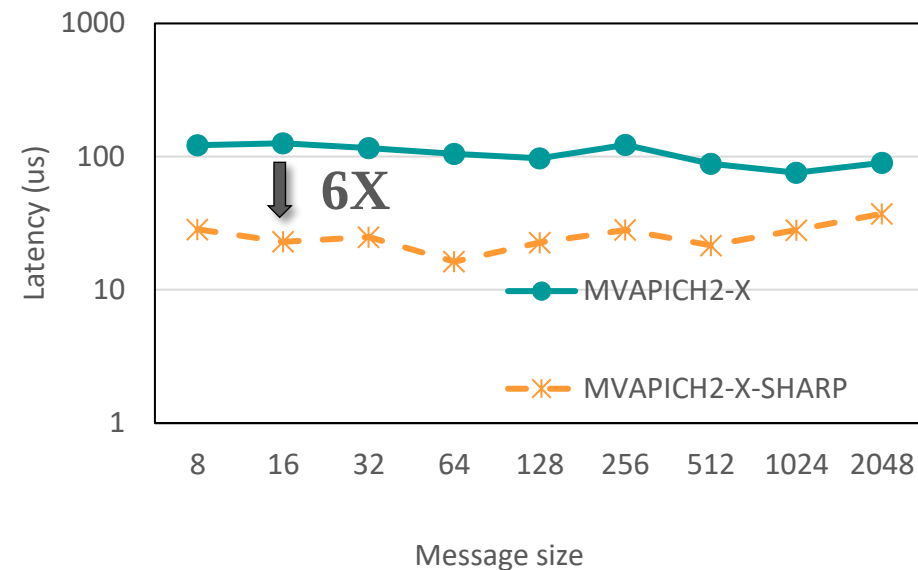
New designs available since MVAPICH2-2.3.4

# Performance of Collectives with SHARP on TACC Frontera

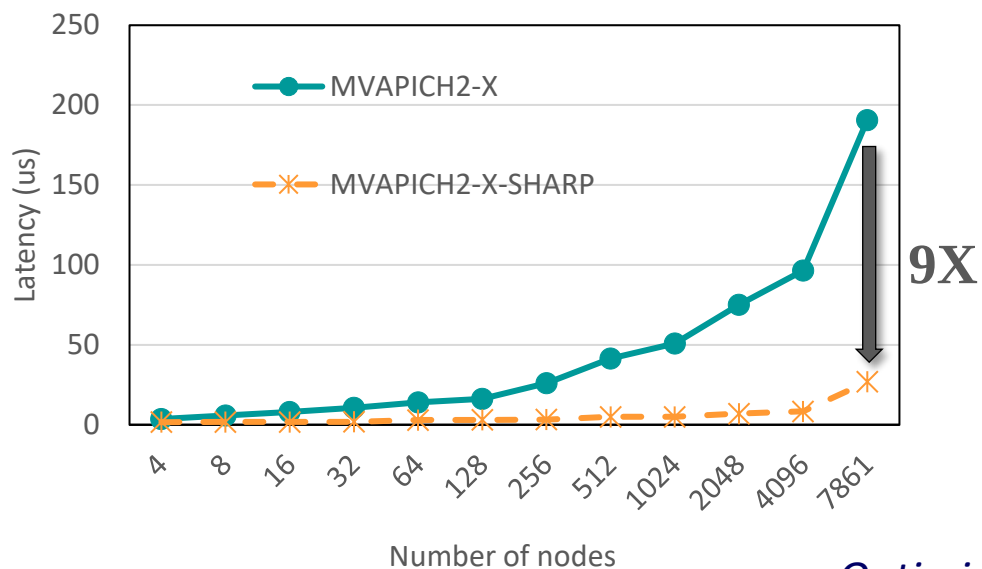
**MPI\_Allreduce**  
(PPN = 1, Nodes = 7861)



**MPI\_Reduce**  
(PPN = 1, Nodes = 7861)



**MPI\_Barrier**



## Optimized SHARP designs in MVAPICH2-X

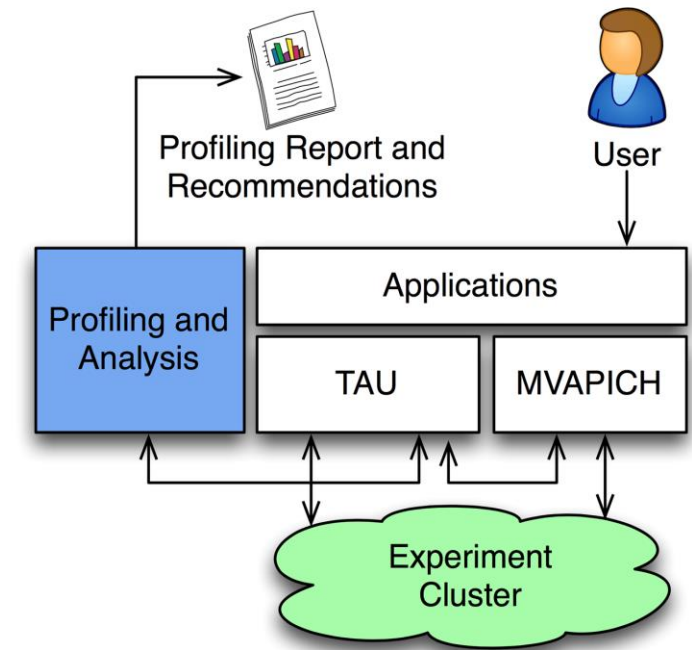
**Up to 9X** performance improvement with SHARP over MVAPICH2-X default for 1ppn MPI\_Barrier, **6X** for 1ppn MPI\_Reduce and **5X** for 1ppn MPI\_Allreduce

B. Ramesh , K. Suresh , N. Sarkauskas , M. Bayatpour , J. Hashmi , H. Subramoni , and D. K. Panda, Scalable MPI Collectives using SHARP: Large Scale Performance Evaluation on the TACC Frontera System, ExaMPI2020 - Workshop on Exascale MPI 2020, Nov 2020.

Optimized Runtime Parameters: `MV2_ENABLE_SHARP = 1`

# Performance Engineering Applications using MVAPICH2 and TAU

- Enhance existing support for MPI\_T in MVAPICH2 to expose a richer set of performance and control variables
- Get and display MPI Performance Variables (PVARs) made available by the runtime in TAU
- Control the runtime's behavior via MPI Control Variables (CVARs)
- Introduced support for new MPI\_T based CVARs to MVAPICH2
  - MPIR\_CVAR\_MAX\_INLINE\_MSG\_SZ, MPIR\_CVAR\_VBUF\_POOL\_SIZE, MPIR\_CVAR\_VBUF\_SECONDARY\_POOL\_SIZE
- TAU enhanced with support for setting MPI\_T CVARs in a non-interactive mode for uninstrumented applications
- S. Ramesh, A. Maheo, S. Shende, A. Malony, H. Subramoni, and D. K. Panda, *MPI Performance Engineering with the MPI Tool Interface: the Integration of MVAPICH and TAU*, *EuroMPI/USA '17, Best Paper Finalist*
- **More details in Sameer Shende's talk (later today)**



**VBUF usage without CVAR based tuning as displayed by ParaProf**

| Name                                                                   | MaxValue  | MinValue  | MeanValue | Std. Dev. | NumSamples | Total     |
|------------------------------------------------------------------------|-----------|-----------|-----------|-----------|------------|-----------|
| mv2_total_vbuf_memory (Total amount of memory in bytes used for VBUFs) | 3,313,056 | 3,313,056 | 3,313,056 | 0         | 1          | 3,313,056 |
| mv2_ud_vbuf_allocated (Number of UD VBUFs allocated)                   | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_ud_vbuf_available (Number of UD VBUFs available)                   | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_ud_vbuf_freed (Number of UD VBUFs freed)                           | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_ud_vbuf_inuse (Number of UD VBUFs inuse)                           | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_ud_vbuf_max_use (Maximum number of UD VBUFs used)                  | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_vbuf_allocated (Number of VBUFs allocated)                         | 320       | 320       | 320       | 0         | 1          | 320       |
| mv2_vbuf_available (Number of VBUFs available)                         | 255       | 255       | 255       | 0         | 1          | 255       |
| mv2_vbuf_freed (Number of VBUFs freed)                                 | 25,545    | 25,545    | 25,545    | 0         | 1          | 25,545    |
| mv2_vbuf_inuse (Number of VBUFs inuse)                                 | 65        | 65        | 65        | 0         | 1          | 65        |
| mv2_vbuf_max_use (Maximum number of VBUFs used)                        | 65        | 65        | 65        | 0         | 1          | 65        |
| num_calloc_calls (Number of MPIT.calloc calls)                         | 89        | 89        | 89        | 0         | 1          | 89        |

**VBUF usage with CVAR based tuning as displayed by ParaProf**

| Name                                                                   | MaxValue  | MinValue  | MeanValue | Std. Dev. | NumSamples | Total     |
|------------------------------------------------------------------------|-----------|-----------|-----------|-----------|------------|-----------|
| mv2_total_vbuf_memory (Total amount of memory in bytes used for VBUFs) | 1,815,056 | 1,815,056 | 1,815,056 | 0         | 1          | 1,815,056 |
| mv2_ud_vbuf_allocated (Number of UD VBUFs allocated)                   | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_ud_vbuf_available (Number of UD VBUFs available)                   | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_ud_vbuf_freed (Number of UD VBUFs freed)                           | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_ud_vbuf_inuse (Number of UD VBUFs inuse)                           | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_ud_vbuf_max_use (Maximum number of UD VBUFs used)                  | 0         | 0         | 0         | 0         | 0          | 0         |
| mv2_vbuf_allocated (Number of VBUFs allocated)                         | 160       | 160       | 160       | 0         | 1          | 160       |
| mv2_vbuf_available (Number of VBUFs available)                         | 94        | 94        | 94        | 0         | 1          | 94        |
| mv2_vbuf_freed (Number of VBUFs freed)                                 | 5,479     | 5,479     | 5,479     | 0         | 1          | 5,479     |
| mv2_vbuf_inuse (Number of VBUFs inuse)                                 | 66        | 66        | 66        | 0         | 1          | 66        |

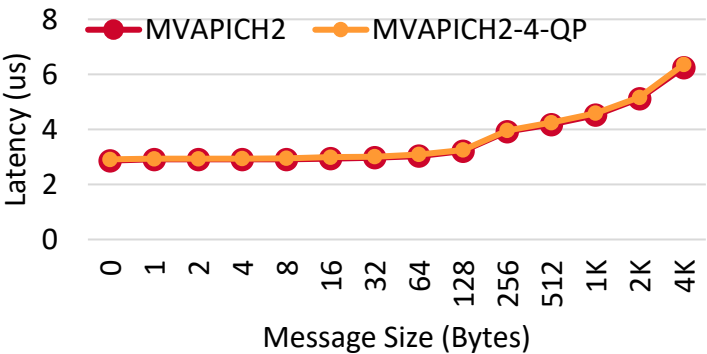
## MVAPICH2 Upcoming Features

- Support for Rockport Networks adapter
- Java Binding for MVAPICH2
- Integration with Sandia SST for large-scale network simulation

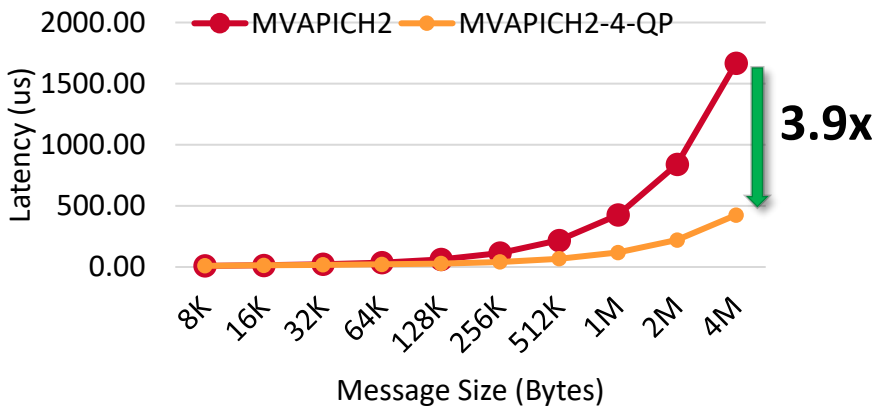


# Inter-node point-to-point Latency and Bandwidth (Rockport Networks)

Small message Latency



Medium/Large message Latency

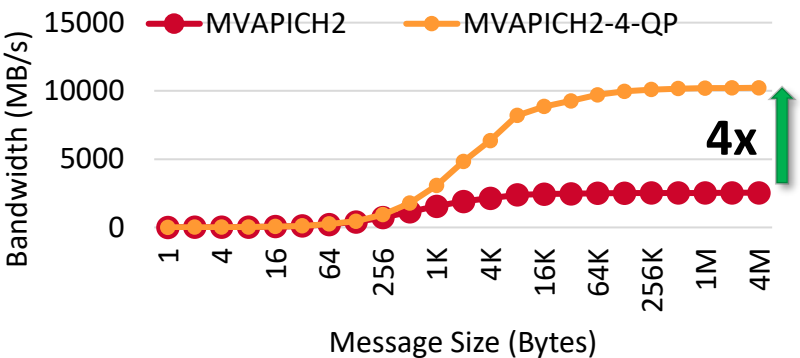


- MVAPICH2 delivers around 3.0 microsec latency for small messages
- Using multiple QPs gives up to 3.9x reduction in latency

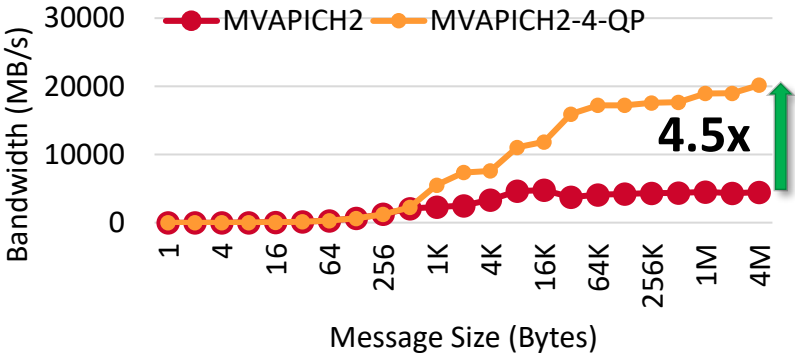
MVAPICH2-4-QP delivers

- 10229 MB/sec peak unidirectional bandwidth
- 20165 MB/Sec peak bi-directional bandwidth

Uni-directional Bandwidth

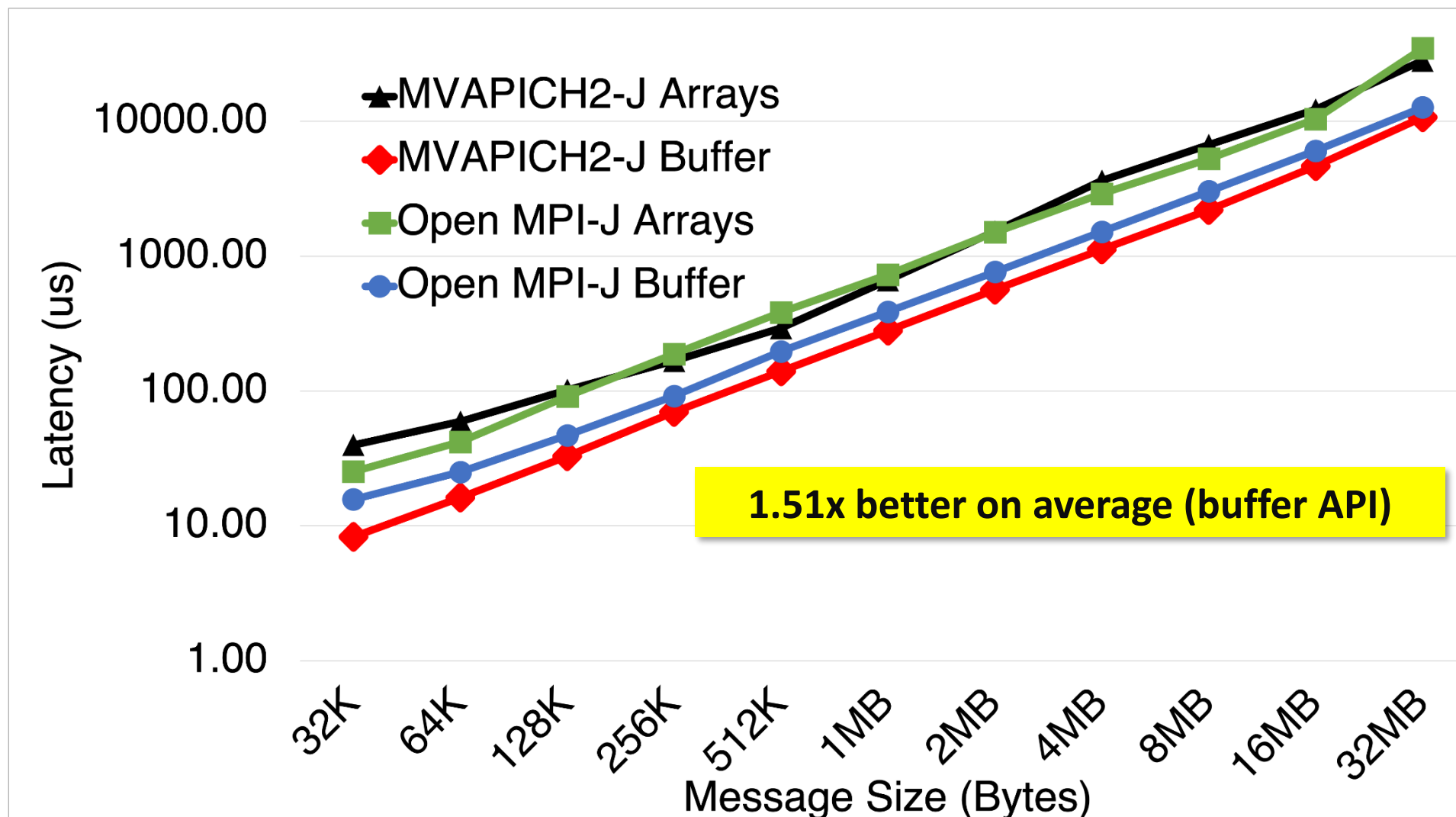


Bi-Directional Bandwidth



More details on today's talk from Rockport Networks

# Java Binding in MVAPICH2: Preliminary Results Bcast Performance (8 processes)



More details in tomorrow's talk by Aamir Shafi

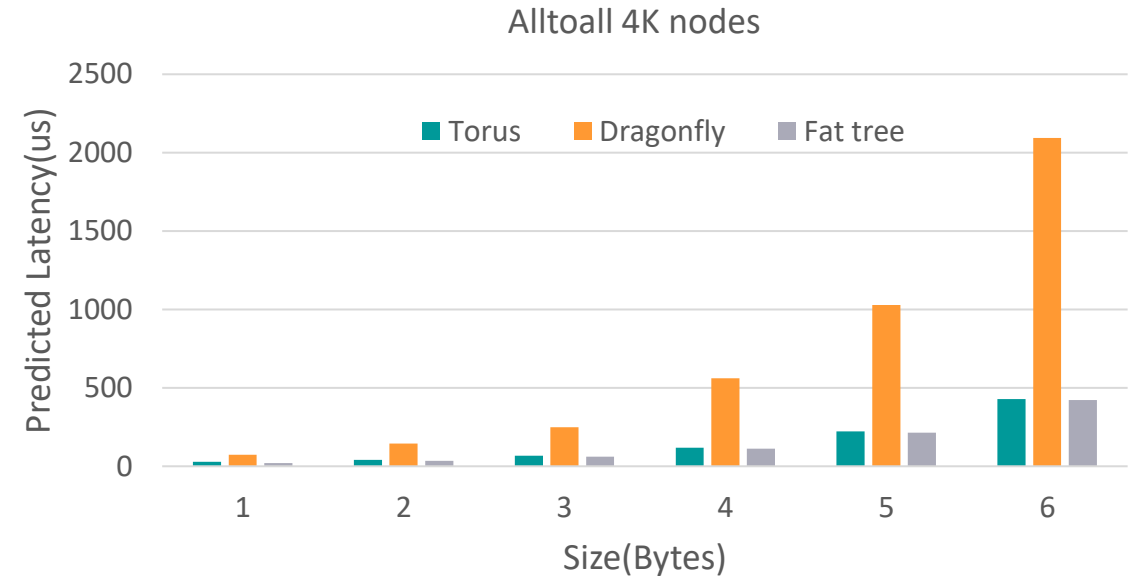
# MVAPICH2 + SST Integration - Understanding the impact of topology

- Large scale simulation of MPI\_Alltoall using OMB
- Impact of topology on Alltoall collective implemented in MVAPICH2

```
topology {  
  name = torus  
  geometry = [32, 32, 32]  
  redundant = [32,32,32]  
}
```

```
topology {  
  name = fat_tree  
  leaf_switches_per_subtree = 64  
  agg_switches_per_subtree = 64  
  concentration = 64  
  up_ports_per_leaf_switch = 64  
  down_ports_per_agg_switch = 64  
  num_agg_subtrees = 2  
  num_core_switches = 64  
  up_ports_per_agg_switch = 64  
  down_ports_per_core_switch = 128  
}
```

```
topology {  
  name = dragonfly  
  geometry = [32,9]  
  h = 16  
  inter_group = circulant  
  concentration = 16  
}
```



More details in tomorrow's talk

By Kaushik Kandadi Suresh

# MVAPICH2 Software Family

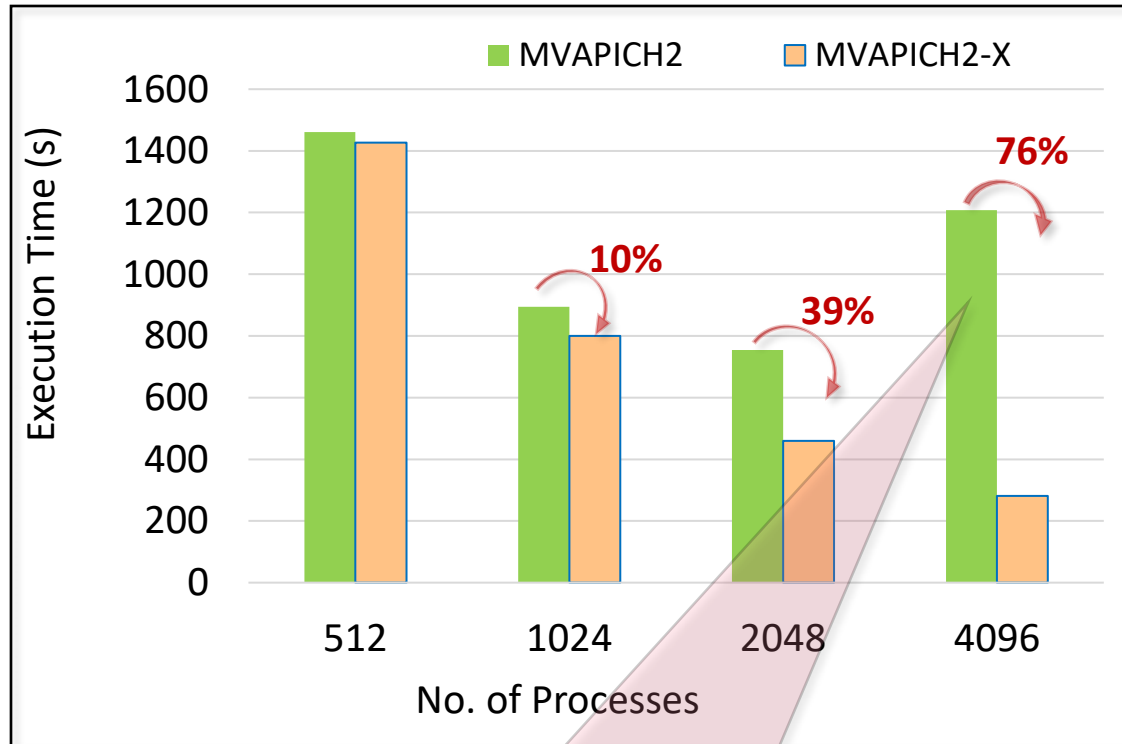
| Requirements                                                                                                                   | Library        |
|--------------------------------------------------------------------------------------------------------------------------------|----------------|
| MPI with Support for InfiniBand, Omni-Path, Ethernet/iWARP and, RoCE (v1/v2)                                                   | MVAPICH2       |
| Optimized Support for Microsoft Azure Platform with InfiniBand                                                                 | MVAPICH2-Azure |
| Advanced MPI features/support (UMR, ODP, DC, Core-Direct, SHArP, XPMEM), OSU INAM (InfiniBand Network Monitoring and Analysis) | MVAPICH2-X     |
| Advanced MPI features (SRD and XPMEM) with support for Amazon Elastic Fabric Adapter (EFA)                                     | MVAPICH2-X-AWS |
| Optimized MPI for clusters with NVIDIA GPUs and for GPU-enabled Deep Learning Applications                                     | MVAPICH2-GDR   |
| Energy-aware MPI with Support for InfiniBand, Omni-Path, Ethernet/iWARP and, RoCE (v1/v2)                                      | MVAPICH2-EA    |
| MPI Energy Monitoring Tool                                                                                                     | OEMT           |
| InfiniBand Network Analysis and Monitoring                                                                                     | OSU INAM       |
| Microbenchmarks for Measuring MPI and PGAS Performance                                                                         | OMB            |

# Overview of Some of the MVAPICH2-X Features

- Direct Connect (DC) Transport
- Cooperative Rendezvous Protocol
- Asynchronous Progress
- XPMEM-based Collectives

# Impact of DC Transport Protocol on Neuron

## Neuron with YuEtAl2012



**Overhead of RC protocol for  
connection establishment and  
communication**

- Up to **76%** benefits over MVAPICH2 for Neuron using Direct Connected transport protocol at scale
  - VERSION 7.6.2 master (f5a1284) 2018-08-15
- Numbers taken on bbpv2.epfl.ch
  - Knights Landing nodes with 64 ppn
  - ./x86\_64/special -mpi -c stop\_time=2000 -c is\_split=1 parinit.hoc
  - Used “runtime” reported by execution to measure performance
- Environment variables used
  - MV2\_USE\_DC=1
  - MV2\_NUM\_DC\_TGT=64
  - MV2\_SMALL\_MSG\_DC\_POOL=96
  - MV2\_LARGE\_MSG\_DC\_POOL=96
  - MV2\_USE\_RDMA\_CM=0

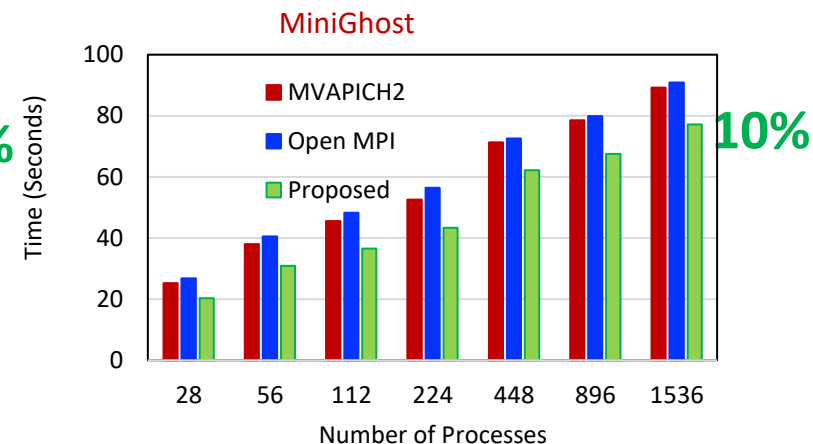
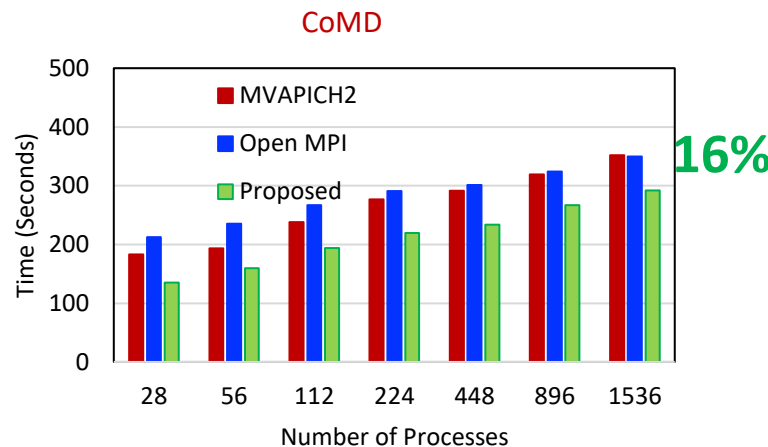
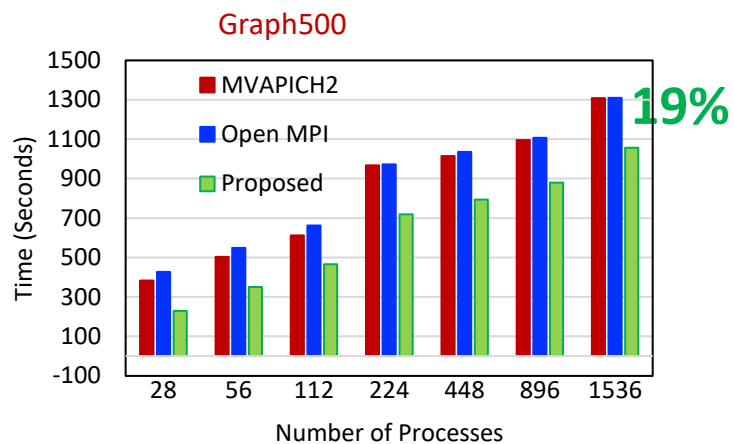
*Available from MVAPICH2-X 2.3rc2 onwards*

*More details in talk*

*“Building Brain Circuits: Experiences with shuffling terabytes of data over MPI”, by Matthias Wolf at MUG’20*

*<https://www.youtube.com/watch?v=TFi8O3-Hznw>*

# Cooperative Rendezvous Protocols



- Use both sender and receiver CPUs to progress communication concurrently
- Dynamically select rendezvous protocol based on communication primitives and sender/receiver availability (load balancing)
- Up to 2x improvement in large message latency and bandwidth
- Up to 19% improvement for Graph500 at 1536 processes

Cooperative Rendezvous Protocols for Improved Performance and Overlap

S. Chakraborty, M. Bayatpour, J Hashmi, H. Subramoni, and DK Panda,

SC '18 (Best Student Paper Award Finalist)

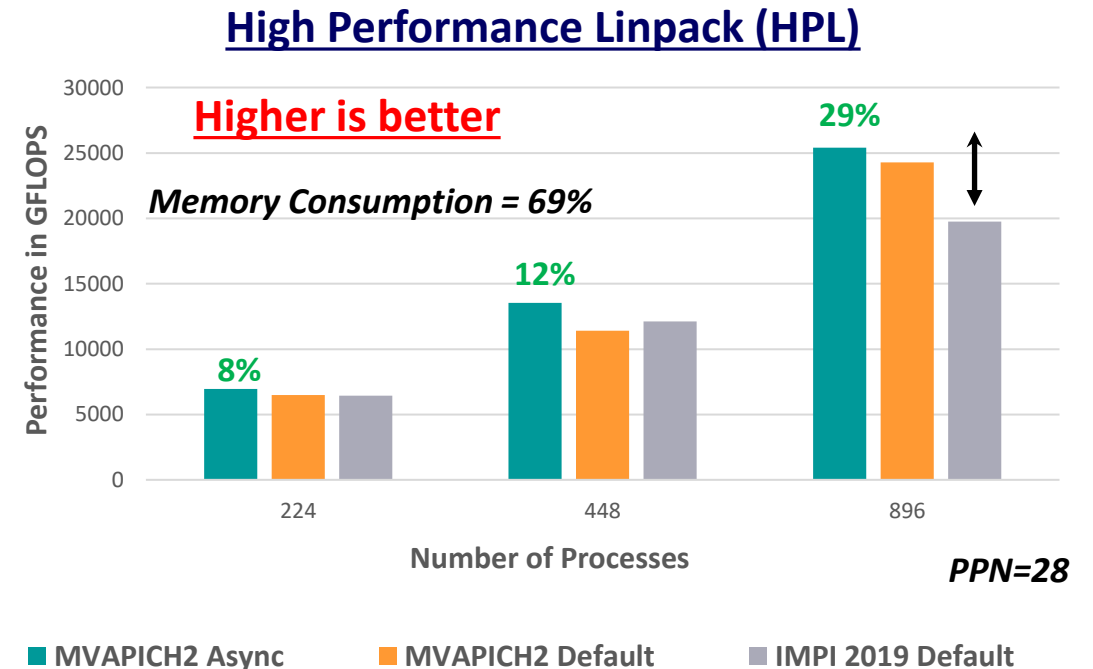
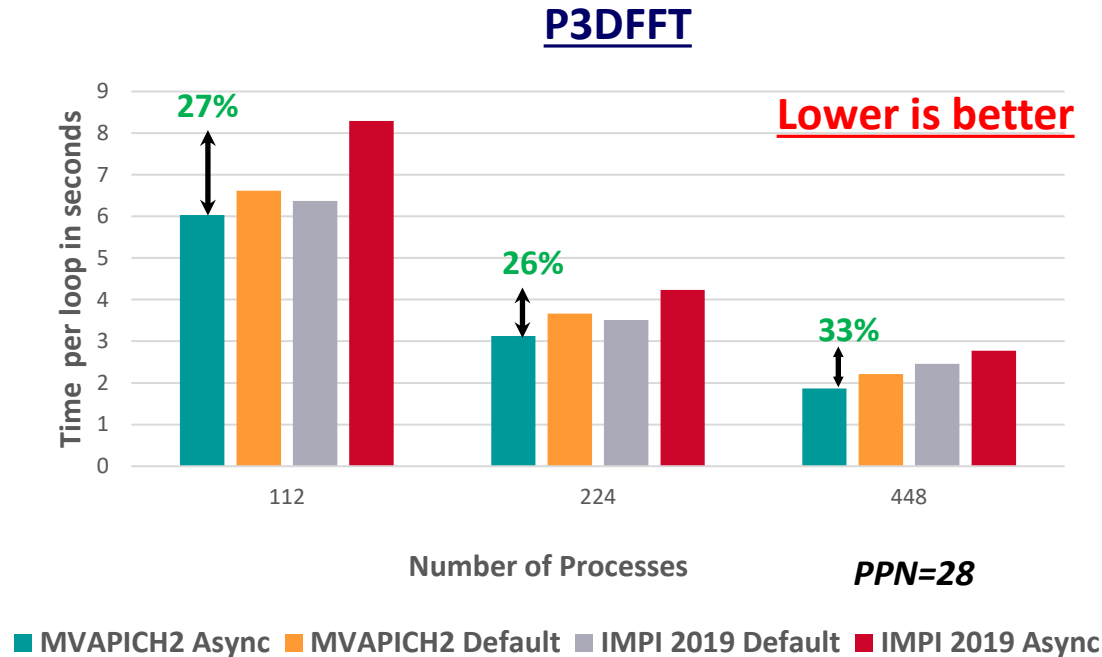
Platform: 2x14 core Broadwell 2680 (2.4 GHz)

Mellanox EDR ConnectX-5 (100 GBps)

Baseline: MVAPICH2X-2.3rc1, Open MPI v3.1.0

Available in MVAPICH2-X 2.3rc2

# Benefits of the New Asynchronous Progress Design: Broadwell + InfiniBand



Up to **33%** performance improvement in P3DFFT application with 448 processes

Up to **29%** performance improvement in HPL application with 896 processes

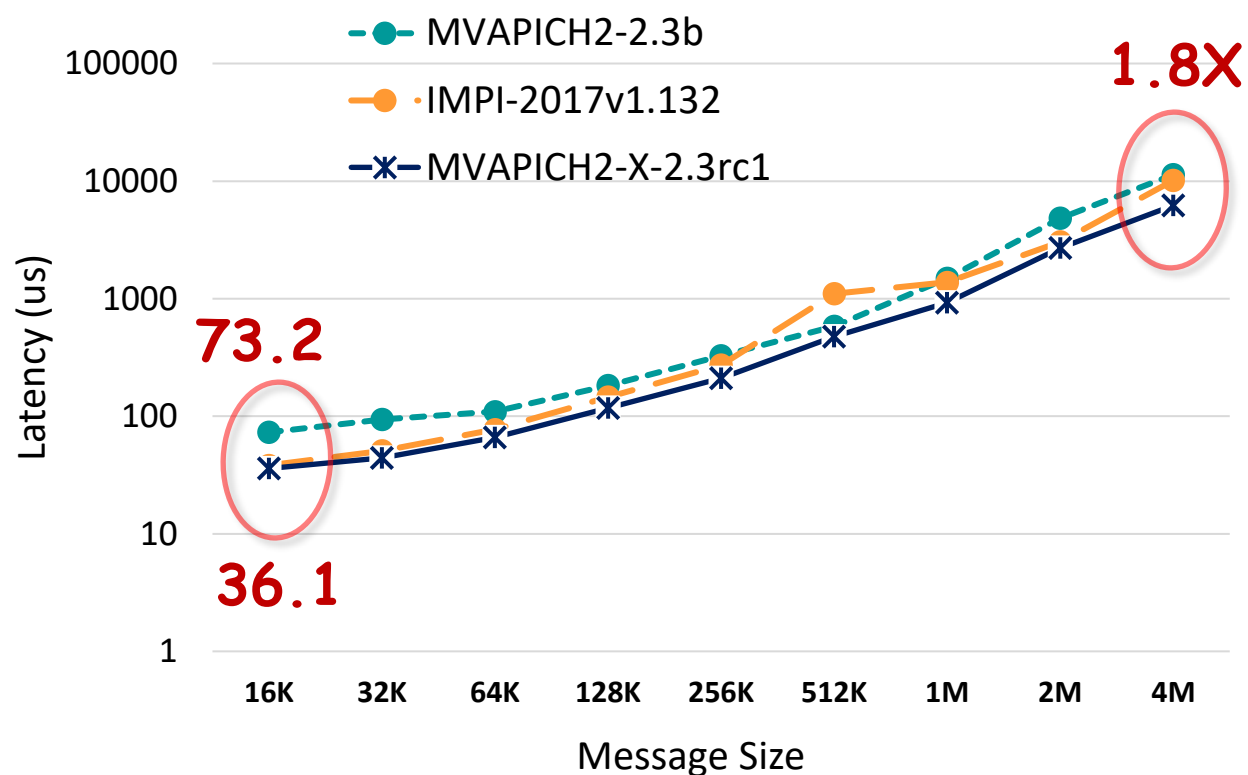
A. Ruhela, H. Subramoni, S. Chakraborty, M. Bayatpour, P. Kousha, and D.K. Panda,  
“Efficient design for MPI Asynchronous Progress without Dedicated Resources”, Parallel Computing 2019

**Available since MVAPICH2-X 2.3rc1**

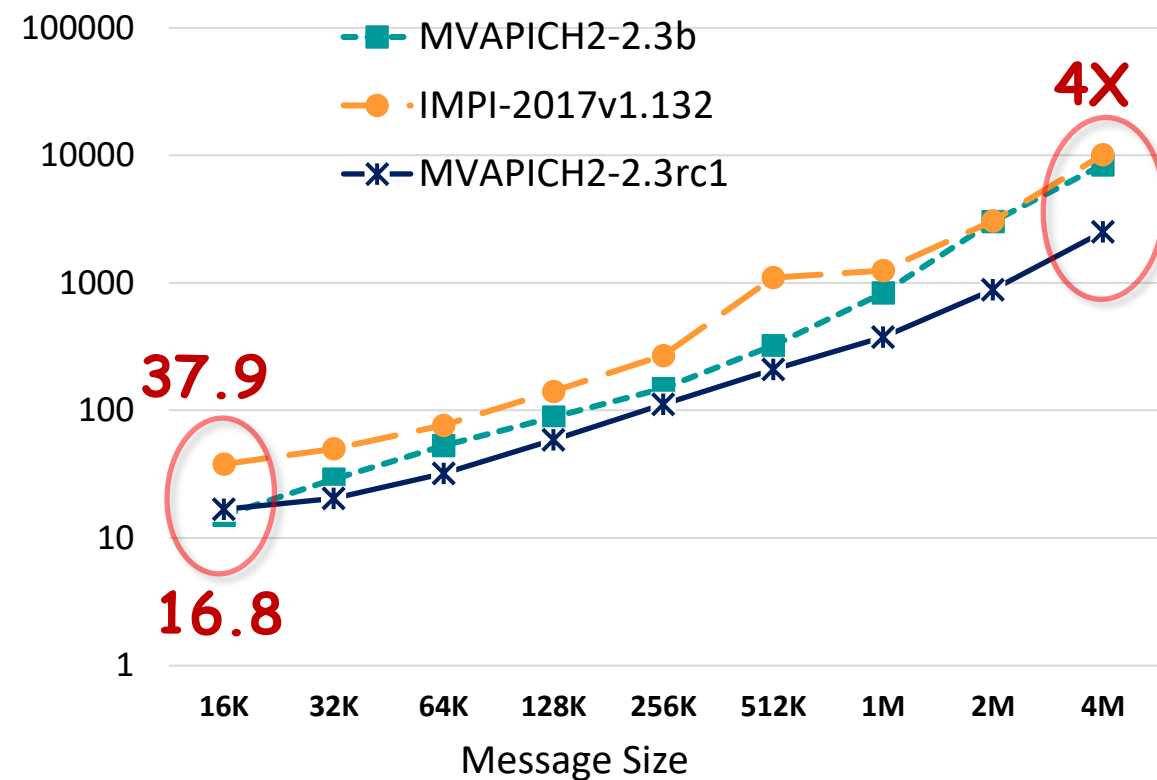


# Shared Address Space (XPMEM)-based Collectives Design

## OSU\_Allreduce (Broadwell 256 procs)



## OSU\_Reduce (Broadwell 256 procs)



- “Shared Address Space”-based true zero-copy Reduction collective designs in MVAPICH2
- Offloaded computation/communication to peers ranks in reduction collective operation
- Up to **4X** improvement for 4MB Reduce and up to **1.8X** improvement for 4M AllReduce

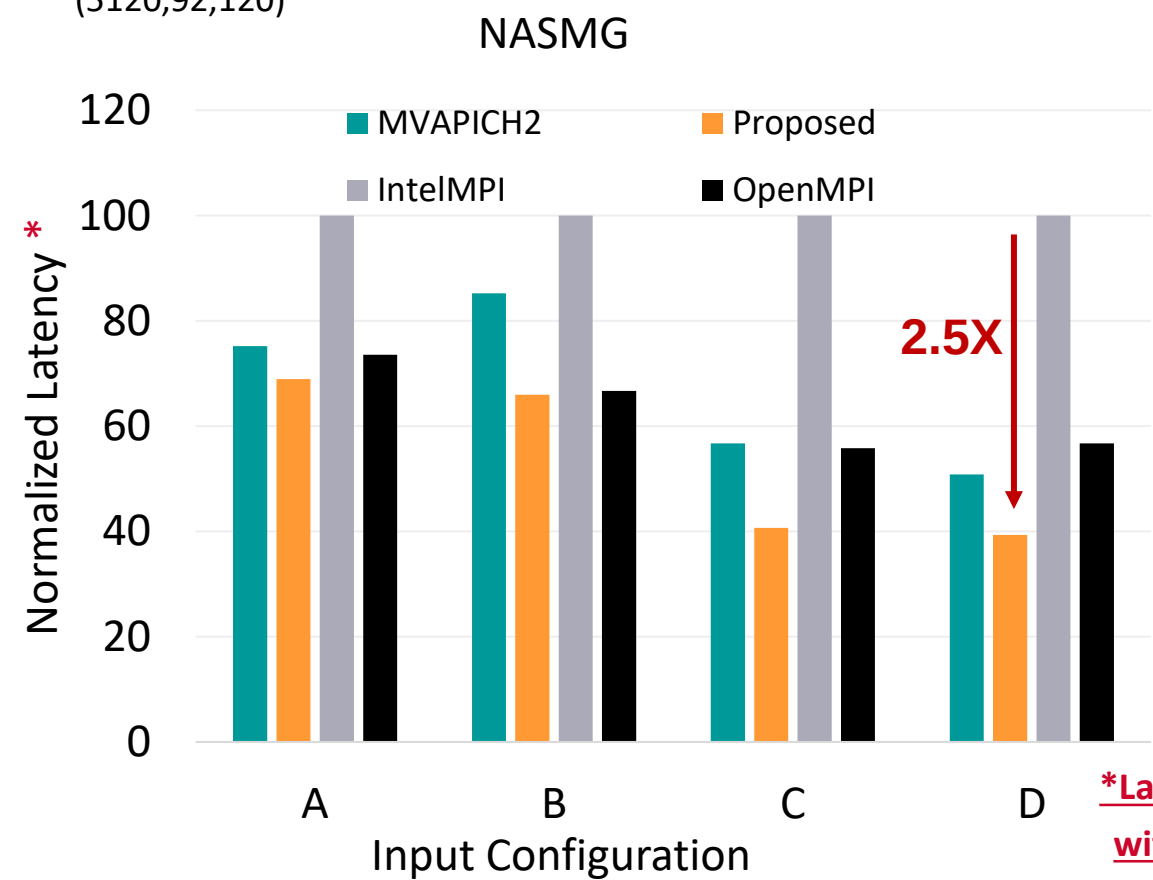
J. Hashmi, S. Chakraborty, M. Bayatpour, H. Subramoni, and D. Panda, *Designing Efficient Shared Address Space Reduction Collectives for Multi-/Many-cores*, International Parallel & Distributed Processing Symposium (IPDPS '18), May 2018. Available since MVAPICH2-X 2.3rc1

# MVAPICH2-X Upcoming Features

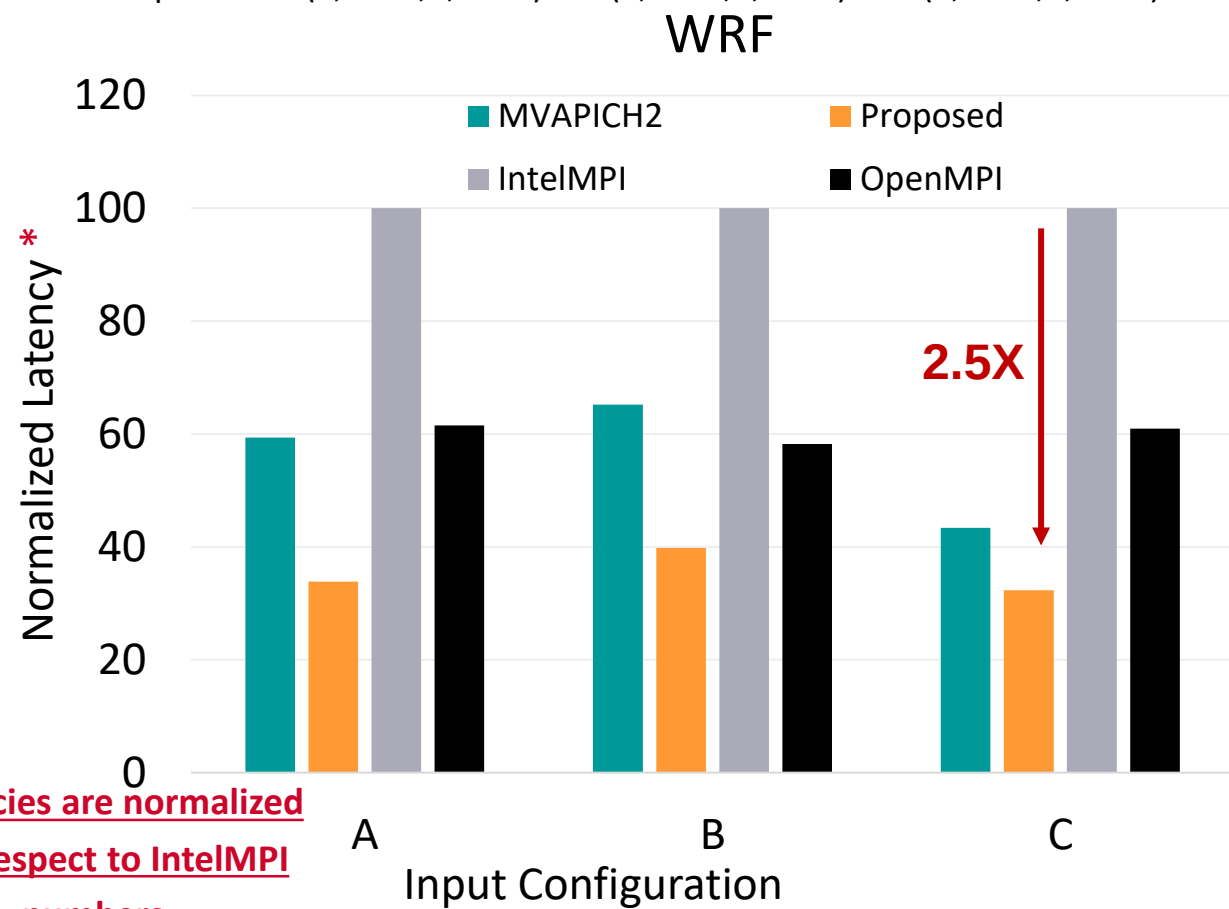
- Optimized Derived Datatype Designs
- Exploiting Hardware Tag Matching
- Neighborhood Collectives
- Support for SHARP Streaming Aggregation (SAT)
- Support for AWS Cloud Arm HPC Instances
- Support for Oracle Cloud HPC Shapes

# Performance of DDTbench with Optimized Derived Datatype Support

- NASMG: Block length is 8 bytes for X-direction and 256 bytes to 5KB in the Y-direction
- 28% improvement over MVAPICH2 and 2.5X over IntelMPI
- Inputs : A = (256,32,32) B = (512,66,66) C = (2048,66,120) D = (5120,92,120)

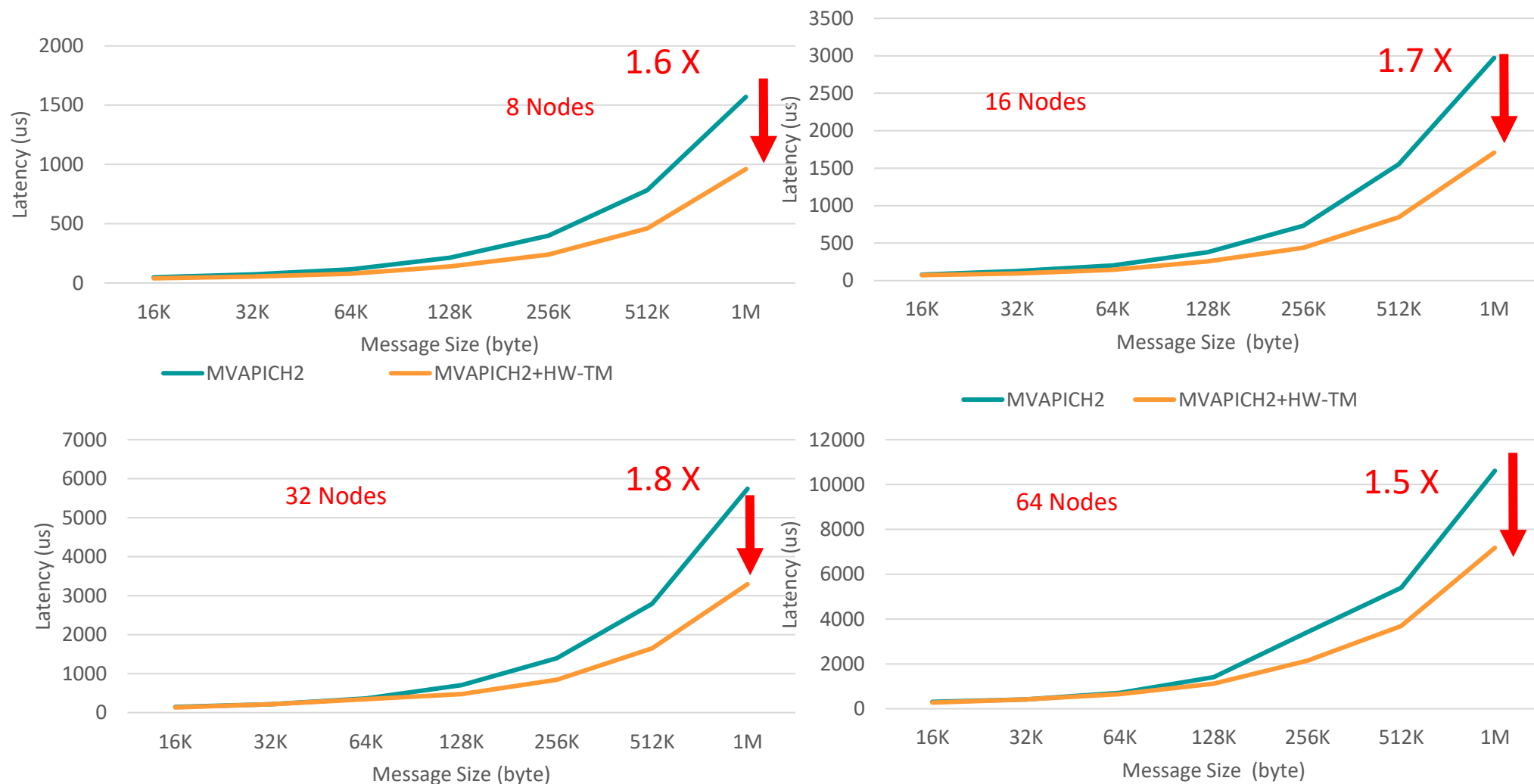


- WRF: The datatypes used in WRF are struct of vectors for both X and Y direction
- We see improvements up to 1.75X compared to MVAPICH2 and up to 2.5X improvements over IntelMPI
- Inputs : A = (4,4018,8,4010) B = (4,2060,8,2056) C = (4,6012,8,6008)



More details in today's talk by Kaushik Kandadi Suresh [numbers](#)

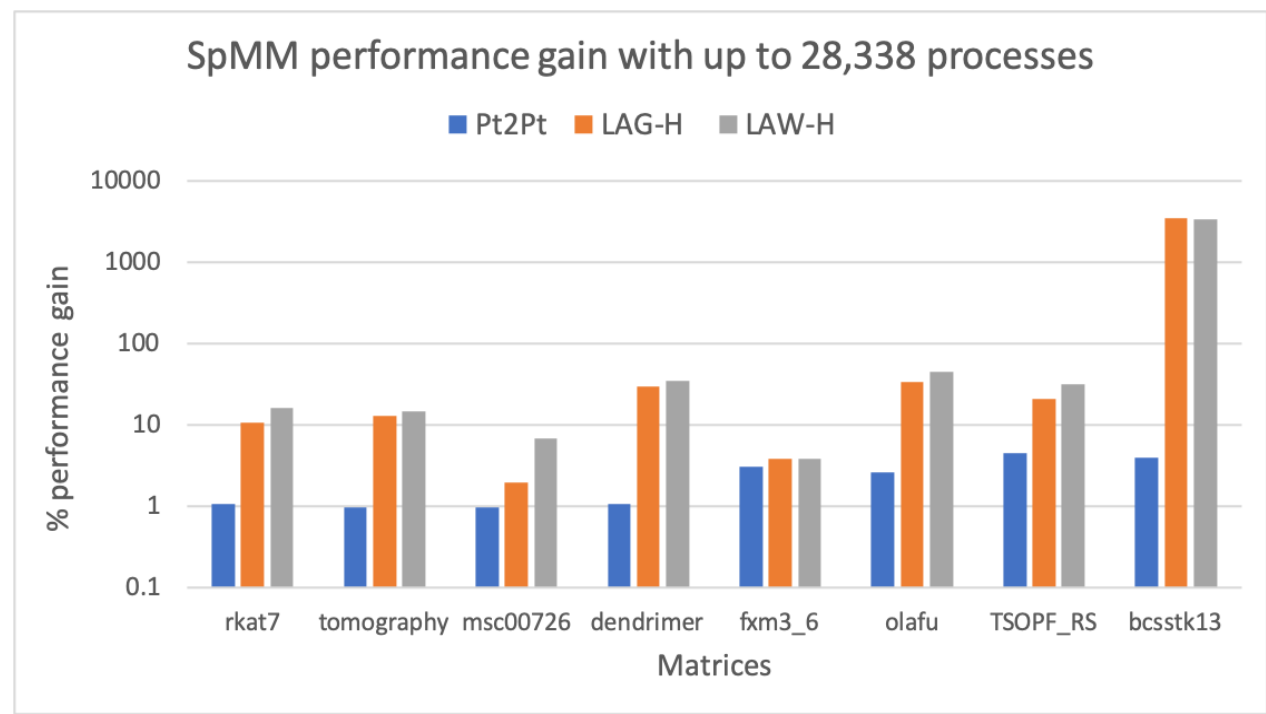
# Performance of MPI\_alltoall using HW Tag Matching



- Up to 1.8x Performance Improvement, Sustained benefits as system size increases

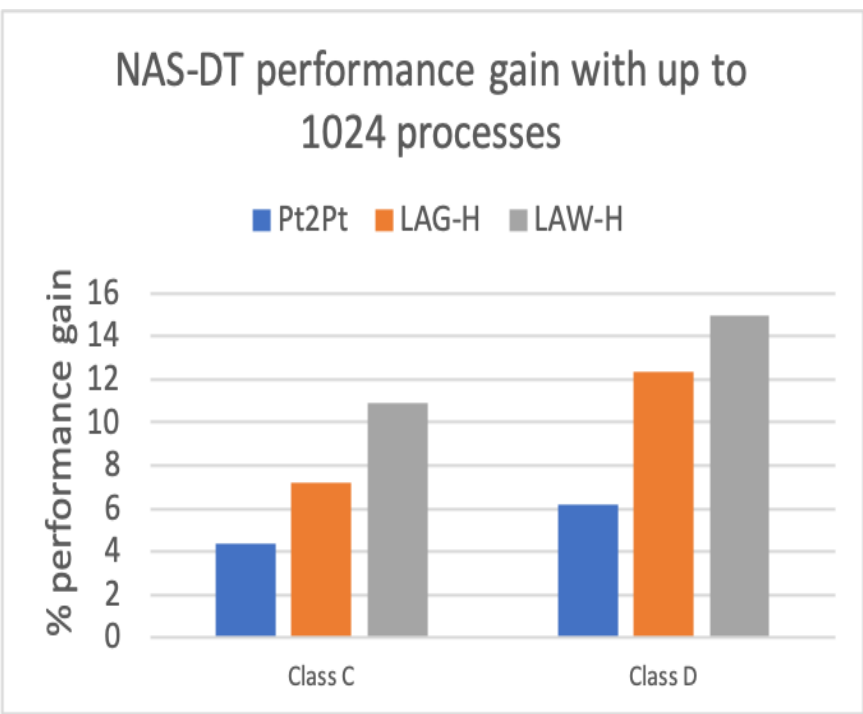
# Optimized Designs for Neighborhood Collectives

- SpMM



up to 34x speedup

- NAS DT

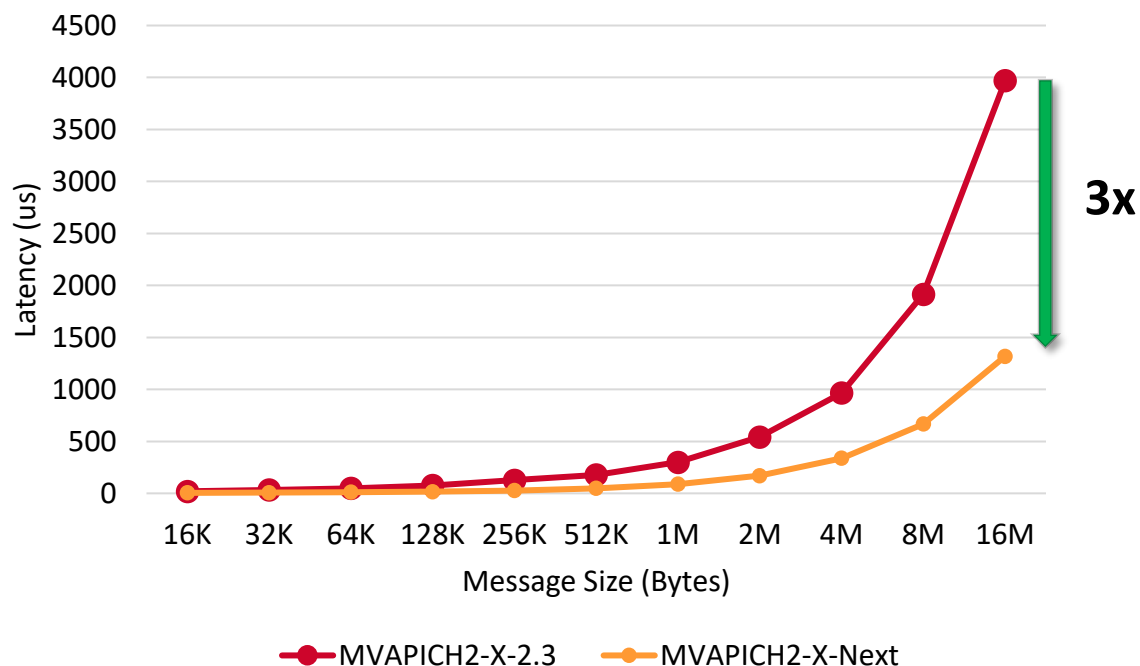


up to 15% improvement

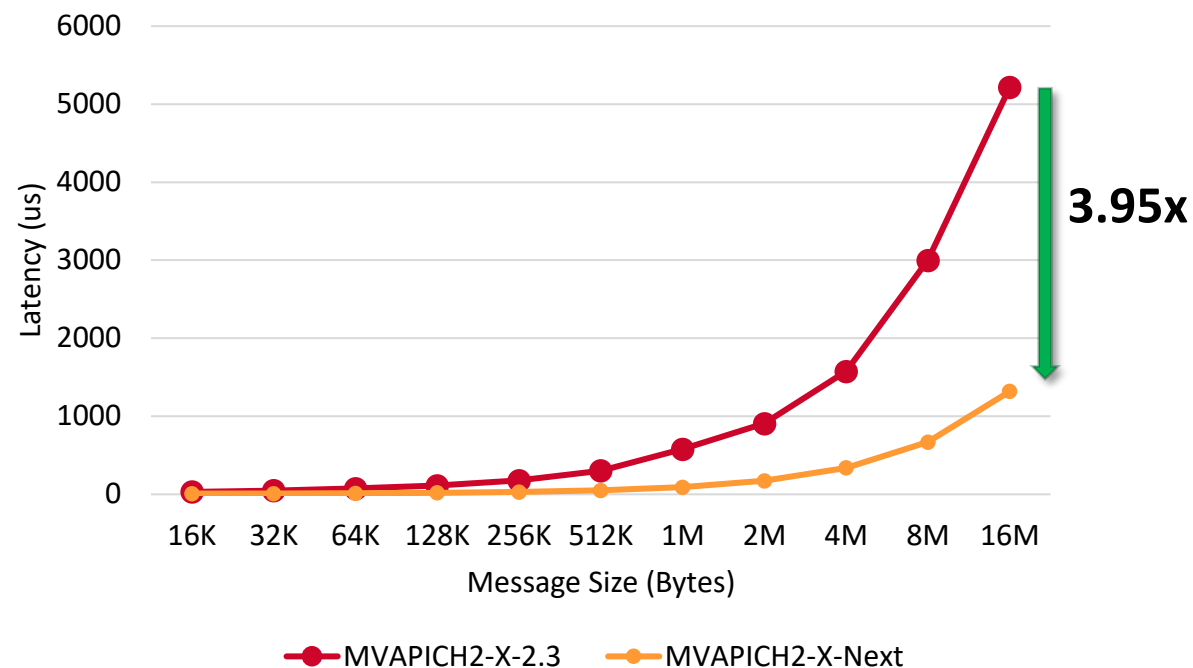
M. Ghazimirsaeed, Q. Zhou, A. Ruhela, M. Bayatpour, H. Subramoni, and DK Panda, “A Hierarchical and Load-Aware Design for Large Message Neighborhood Collectives”, SC ‘20

# Optimized MPI\_Allreduce Performance with MVAPICH2-X + SHARP SAT

4 nodes, 1 PPN



16 nodes, 1 PPN

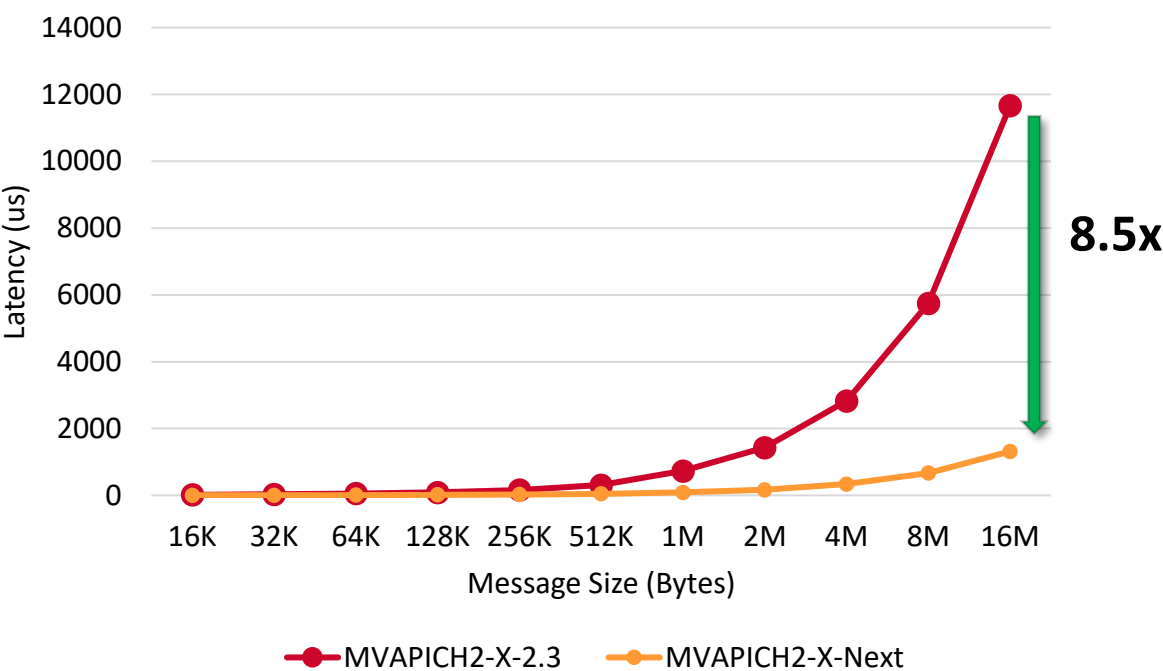


- SHARP provides flat scaling, even for large messages
- Up to **3.95X** benefits over MVAPICH2-X-2.3 using SAT + optimized designs

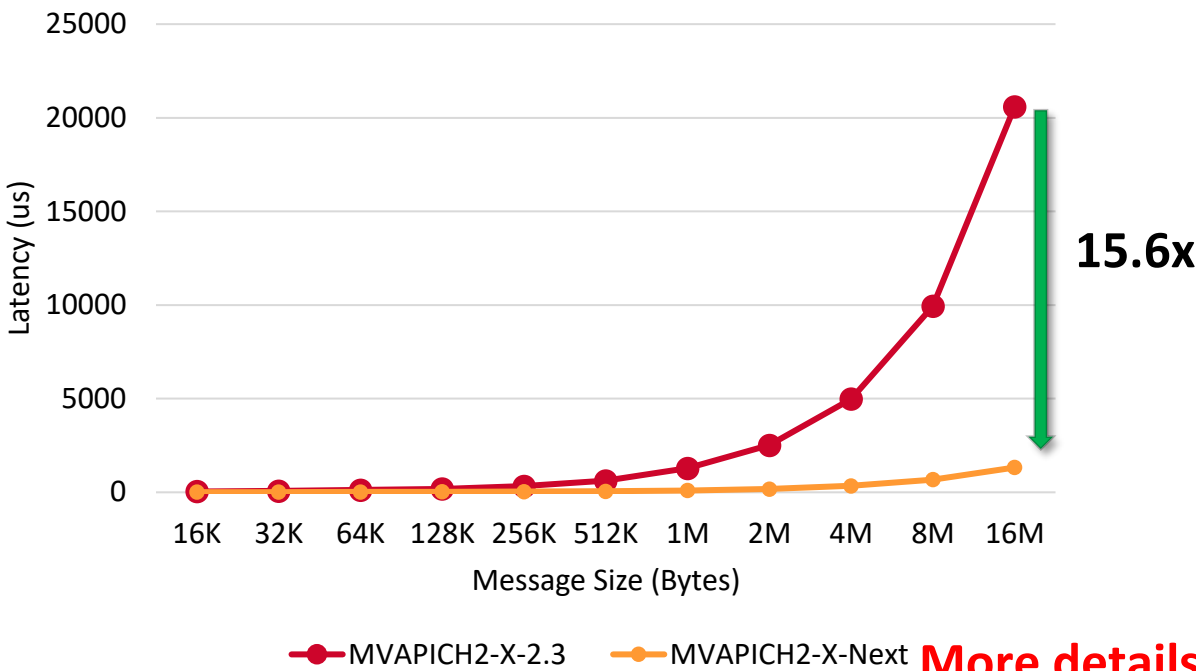
Platform: Intel(R) Xeon(R) Gold 6138 nodes equipped with a dual-socket CPU and InfiniBand HDR-200 Interconnect

# Optimized MPI\_Reduce Performance with MVAPICH2-X + SHARP SAT

4 nodes, 1 PPN



16 nodes, 1 PPN

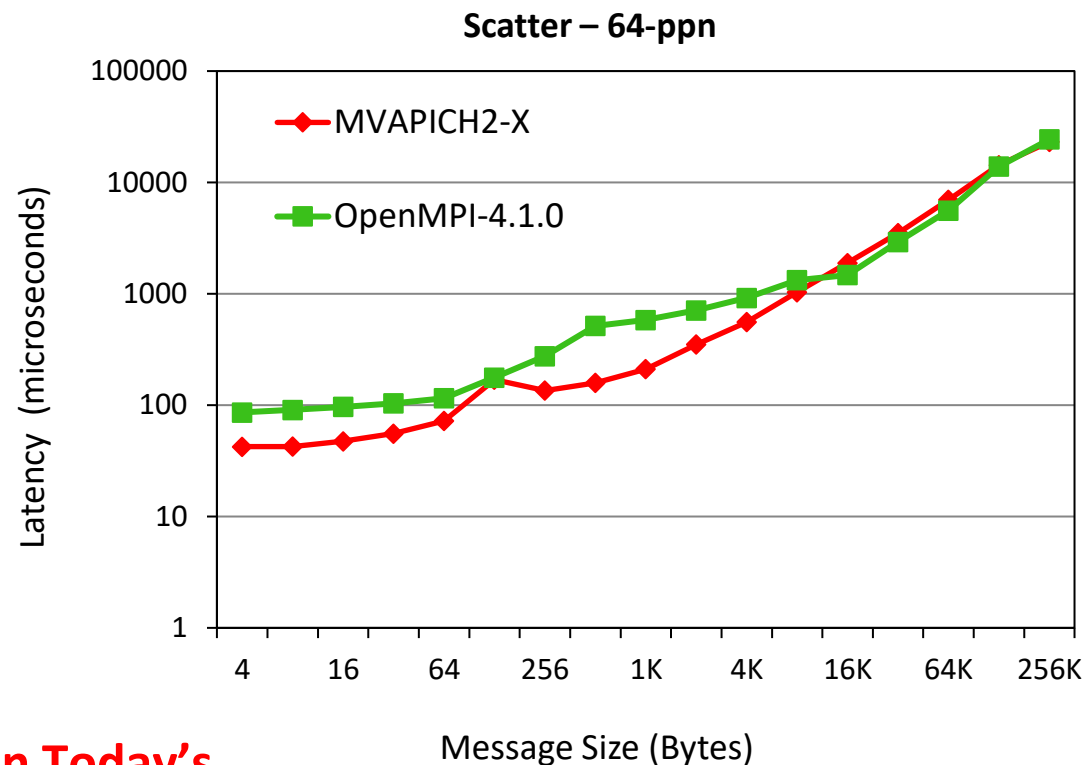
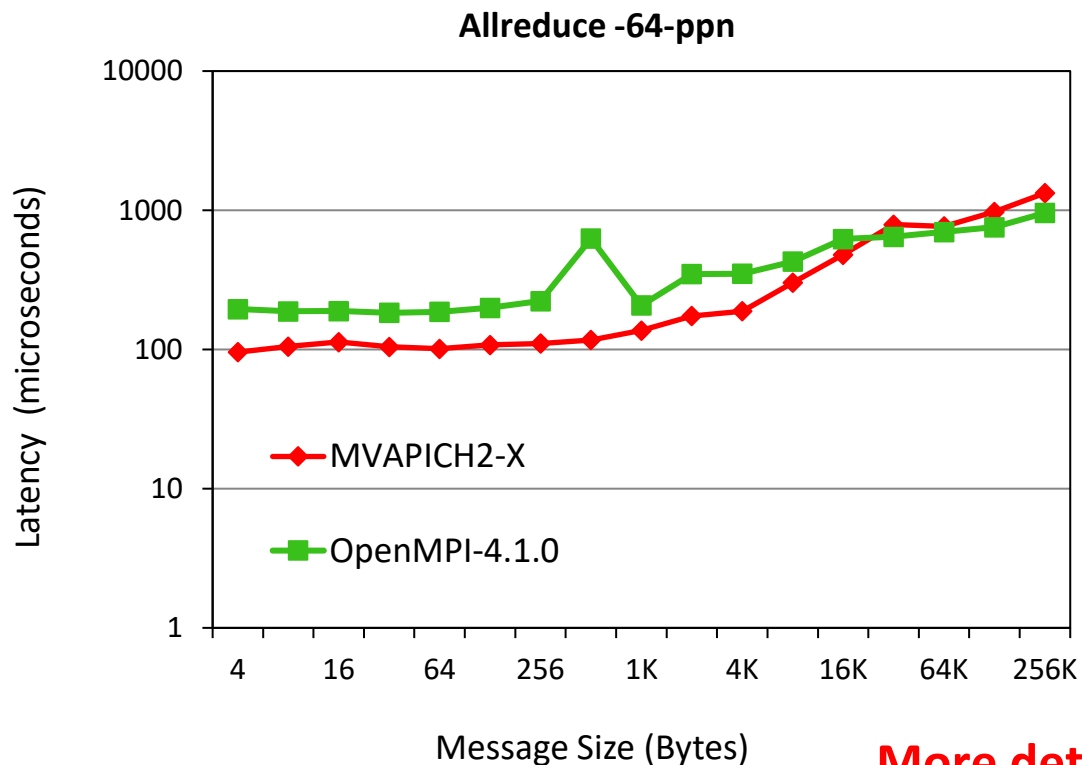


- Comparing max latency for MPI\_Reduce as the root is the bottleneck
  - Up to **15.6X** benefits over MVAPICH2-X-2.3 using SAT + optimized designs
- More details in Today's talk by Bharath Ramesh

Platform: Intel(R) Xeon(R) Gold 6138 nodes equipped with a dual-socket CPU and InfiniBand HDR-200 Interconnect

# MVAPICH2-X on AWS EFA Arm HPC Instances

- Collective Performance on 32 AWS c6gn.18xlarge instances

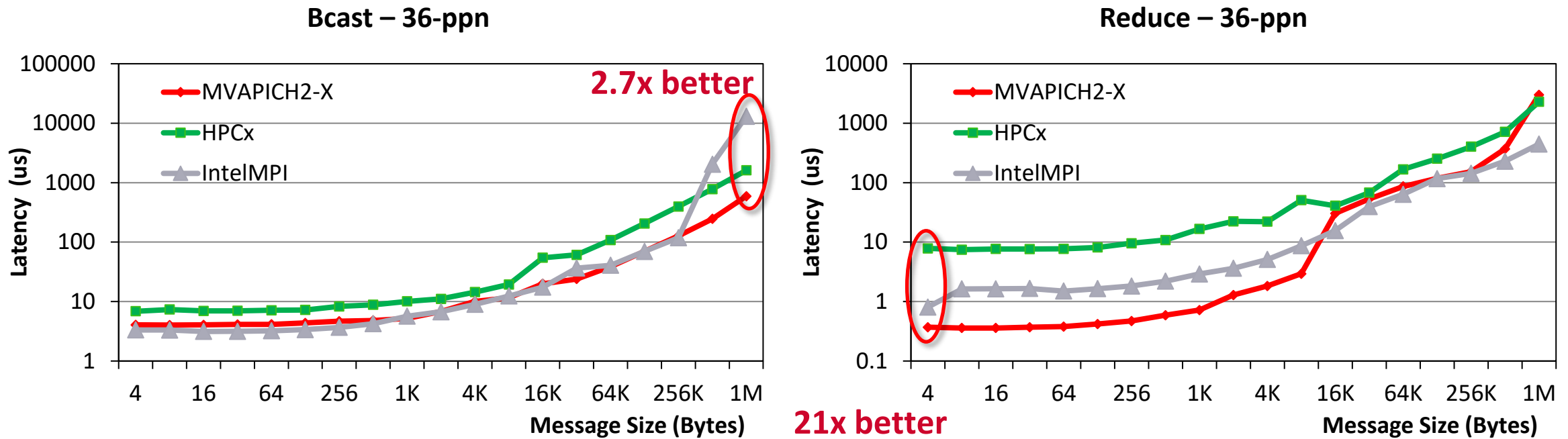


**More details in Today's  
Talk by Shulei Xu**



# MVAPICH2-X Performance on OCI HPC System

- Collective performance evaluation on 8 BM.HPC2 instances



More details in Today's  
Talk by Shulei Xu

# MVAPICH2 Software Family

| Requirements                                                                                                                   | Library             |
|--------------------------------------------------------------------------------------------------------------------------------|---------------------|
| MPI with Support for InfiniBand, Omni-Path, Ethernet/iWARP and, RoCE (v1/v2)                                                   | MVAPICH2            |
| Optimized Support for Microsoft Azure Platform with InfiniBand                                                                 | MVAPICH2-Azure      |
| Advanced MPI features/support (UMR, ODP, DC, Core-Direct, SHArP, XPMEM), OSU INAM (InfiniBand Network Monitoring and Analysis) | MVAPICH2-X          |
| Advanced MPI features (SRD and XPMEM) with support for Amazon Elastic Fabric Adapter (EFA)                                     | MVAPICH2-X-AWS      |
| <b>Optimized MPI for clusters with NVIDIA GPUs and for GPU-enabled Deep Learning Applications</b>                              | <b>MVAPICH2-GDR</b> |
| Energy-aware MPI with Support for InfiniBand, Omni-Path, Ethernet/iWARP and, RoCE (v1/v2)                                      | MVAPICH2-EA         |
| MPI Energy Monitoring Tool                                                                                                     | OEMT                |
| InfiniBand Network Analysis and Monitoring                                                                                     | OSU INAM            |
| Microbenchmarks for Measuring MPI and PGAS Performance                                                                         | OMB                 |

# MVAPICH2-GDR 2.3.6

- Released on 08/12/2021
- Major Features and Enhancements
  - Based on MVAPICH2 2.3.6
  - Added support for 'on-the-fly' compression of point-to-point messages used for GPU-to-GPU communication
    - Applicable to NVIDIA GPUs
  - Added NCCL communication substrate for various MPI collectives
    - Support for hybrid communication protocols using NCCL-based, CUDA-based, and IB verbs-based primitives
    - MPI\_Allreduce, MPI\_Reduce, MPI\_Allgather, MPI\_Allgatherv, MPI\_Alltoall, MPI\_Alltoallv, MPI\_Scatter, MPI\_Scatterv, MPI\_Gather, MPI\_Gatherv, and MPI\_Bcast
  - Full support for NVIDIA DGX, NVIDIA DGX-2 V-100, and NVIDIA DGX-2 A-100 systems
    - Enhanced architecture detection, process placement and HCA selection
    - Enhanced intra-node and inter-node point-to-point tuning
    - Enhanced collective tuning
  - Introduced architecture detection, point-to-point tuning and collective tuning for ThetaGPU @ANL
  - Enhanced point-to-point and collective tuning for NVIDIA GPUs on Frontera @TACC, Lassen @LLNL, and Sierra @LLNL
  - Enhanced point-to-point and collective tuning for Mi50 and Mi60 AMD GPUs on Corona @LLNL
  - Added several new MPI\_T PVARs
  - Added support for CUDA 11.3
  - Added support for ROCm 4.1
  - Enhanced output for runtime variable MV2\_SHOW\_ENV\_INFO
  - Tested with Horovod and common DL Frameworks
    - TensorFlow, PyTorch, and MXNet
  - Tested with MPI4Dask 0.2
    - MPI4Dask is a custom Dask Distributed package with MPI support
  - Tested with MPI4cuML 0.1
    - MPI4cuML is a custom cuML package with MPI support

# Highlights of some MVAPICH2-GDR Features for HPC and DL

- CUDA-Aware MPI
- Support for AMD GPU
- On-the-fly Compression for GPU-GPU Communication
- Optimized Collective Support for DGX-A100
- High-Performance
  - Deep Learning
  - Machine Learning
  - Data Science with Dask

# 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

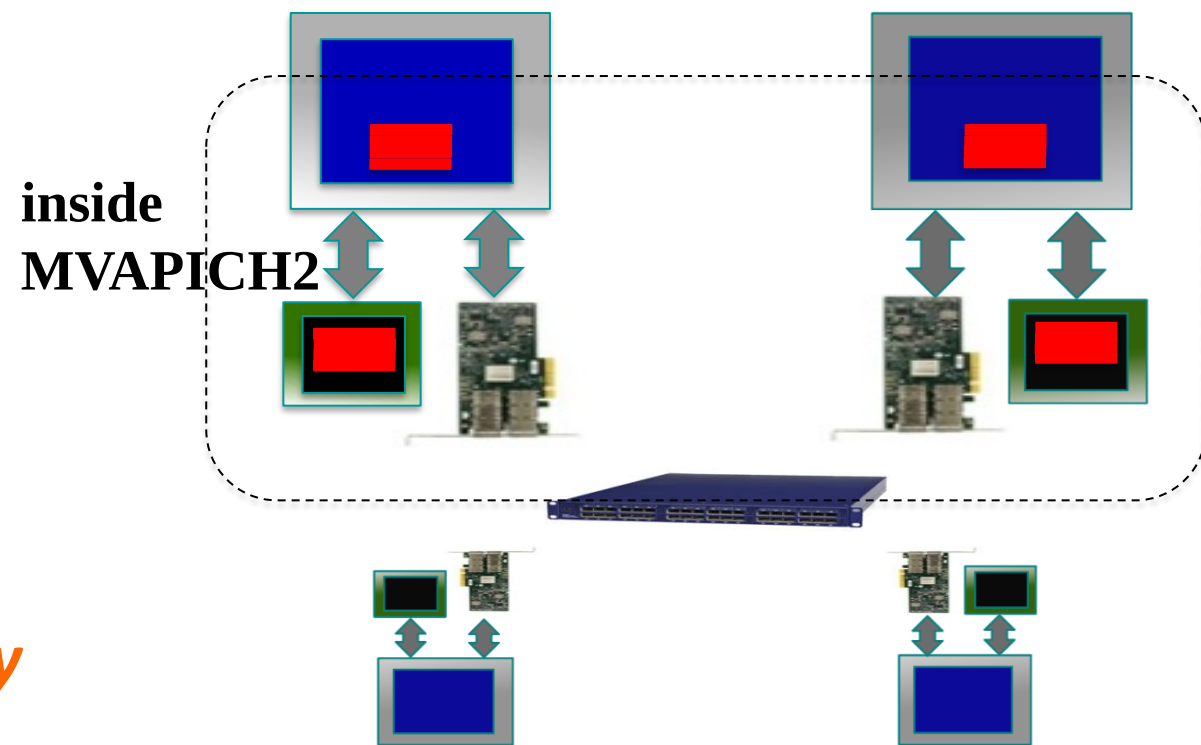
## At Sender:

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

## At Receiver:

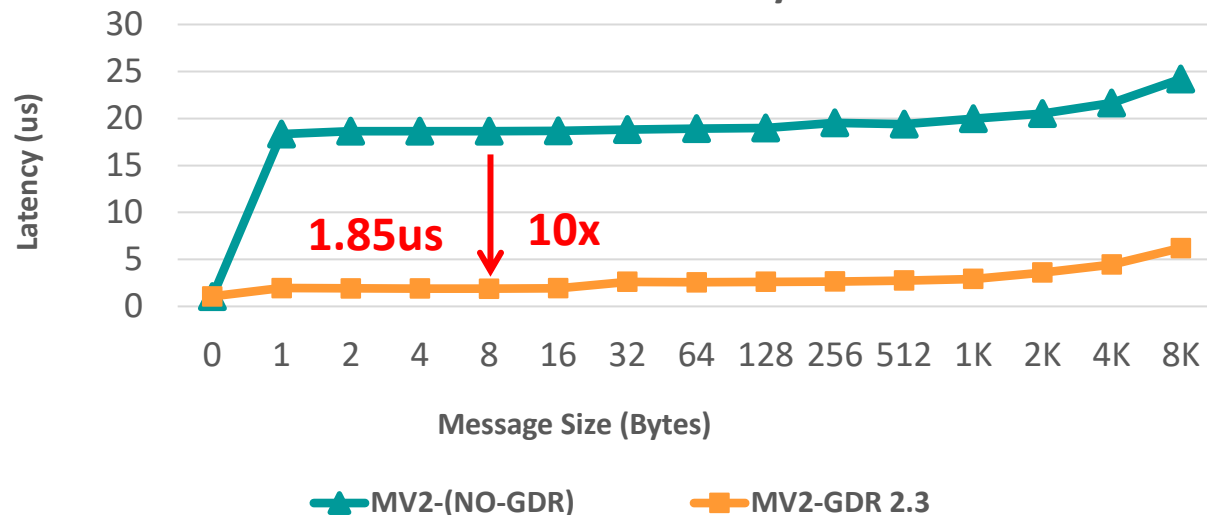
```
MPI_Recv(r_devbuf, size, ...);
```

*High Performance and High Productivity*

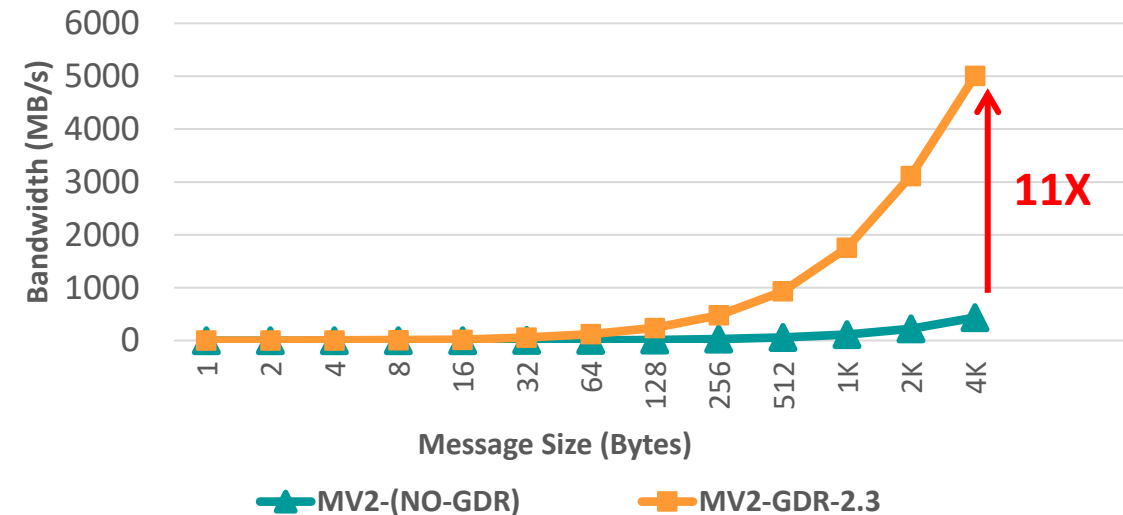


# MVAPICH2-GDR with CUDA-aware MPI Support

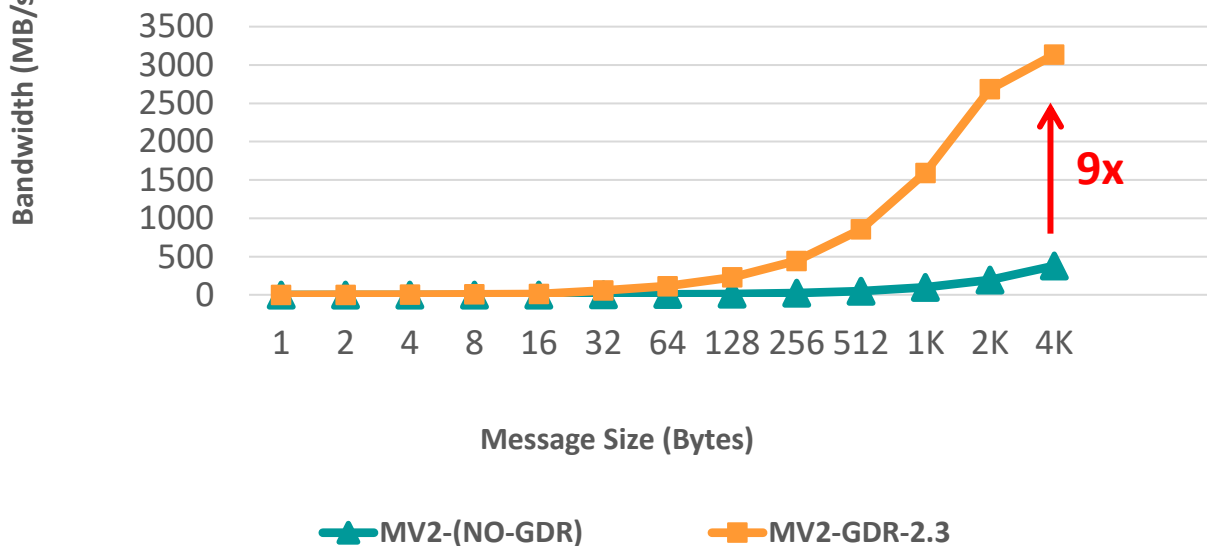
## GPU-GPU Inter-node Latency



## GPU-GPU Inter-node Bi-Bandwidth

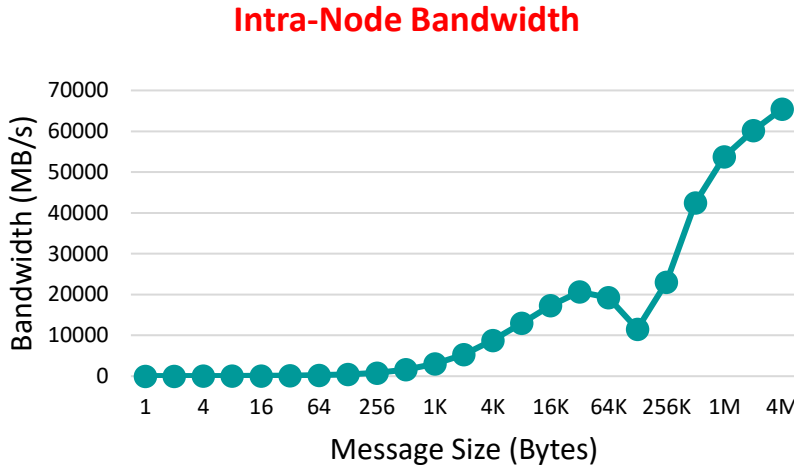
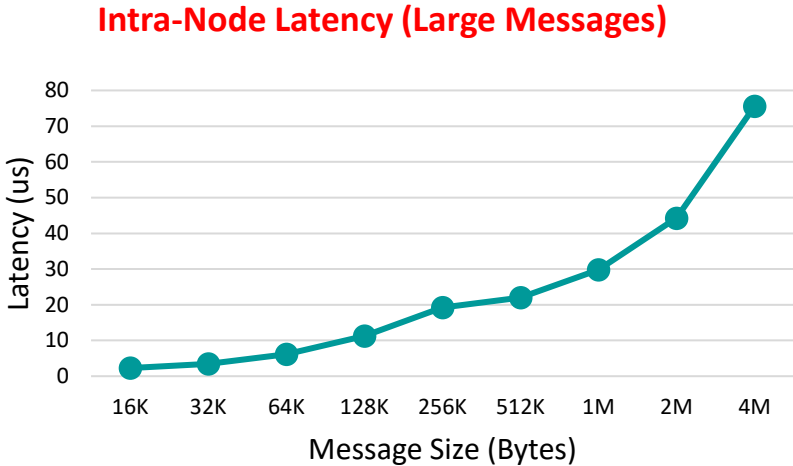
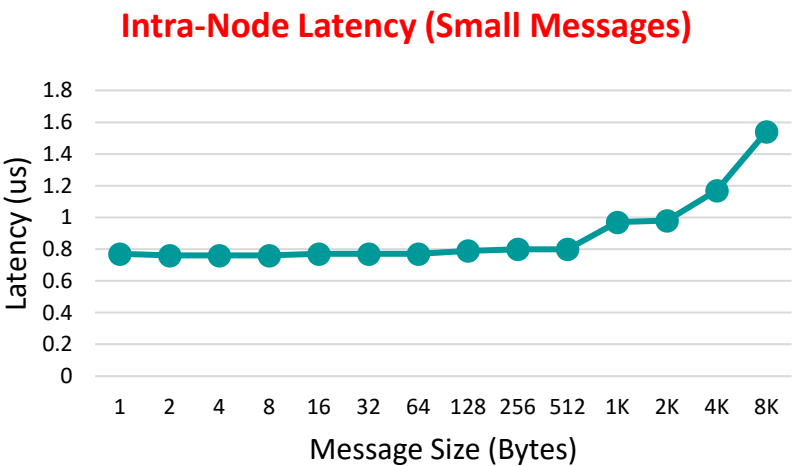


## GPU-GPU Inter-node Bandwidth



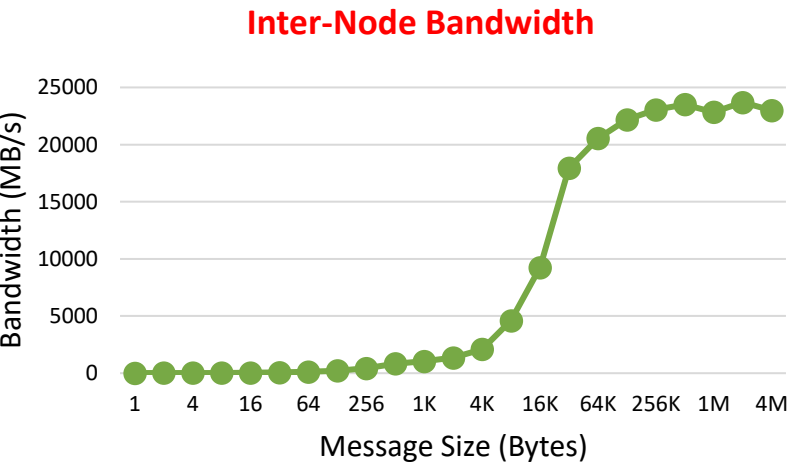
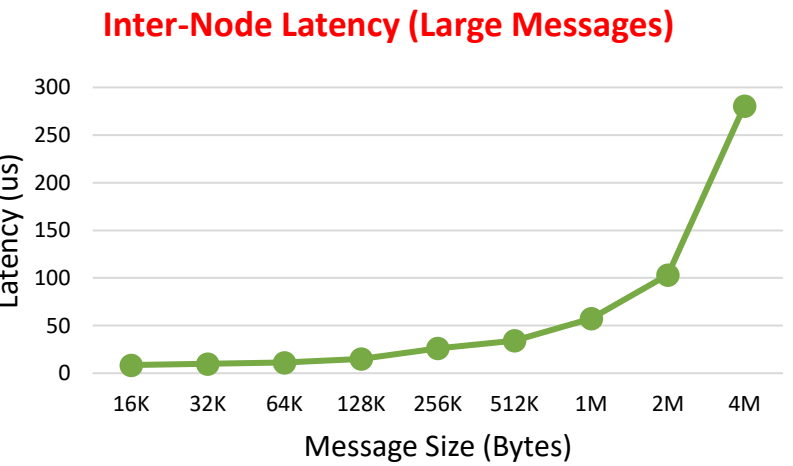
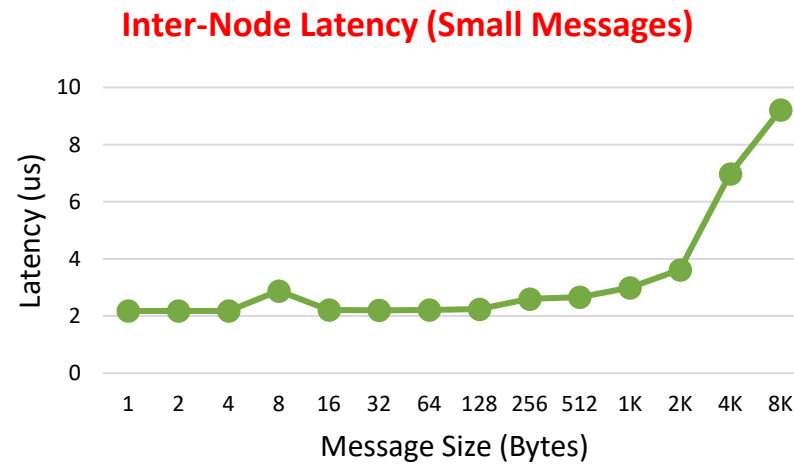
**MVAPICH2-GDR-2.3**  
**Intel Haswell (E5-2687W @ 3.10 GHz) node - 20 cores**  
**NVIDIA Volta V100 GPU**  
**Mellanox Connect-X4 EDR HCA**  
**CUDA 9.0**  
**Mellanox OFED 4.0 with GPU-Direct-RDMA**

# D-to-D Performance on OpenPOWER w/ GDRCopy (NVLink2 + Volta)



*Intra-node Latency: 0.76 us (with GDRCopy)*

*Intra-node Bandwidth: 65.48 GB/sec for 4MB (via NVLINK2)*



*Inter-node Latency: 2.18 us (with GDRCopy 2.0)*

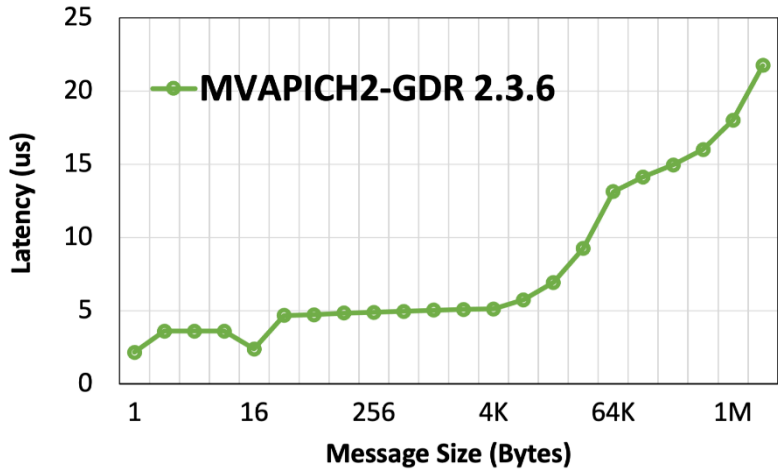
*Inter-node Bandwidth: 23 GB/sec for 4MB (via 2 Port EDR)*

*Platform: OpenPOWER (POWER9-ppc64le) nodes equipped with a dual-socket CPU, 4 Volta V100 GPUs, and 2port EDR InfiniBand Interconnect*

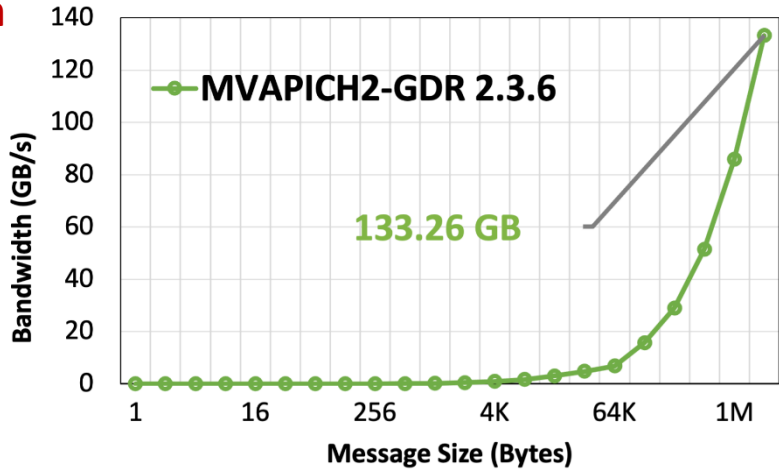
# AMD ND A100 v4-series (NDv4) + (8 x HDR 200)

## Intra-Node GPU Point-to-Point

Latency

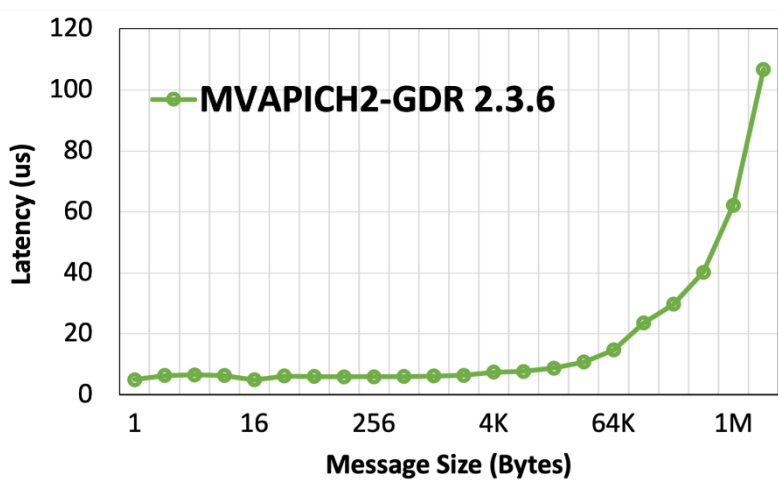


Bandwidth

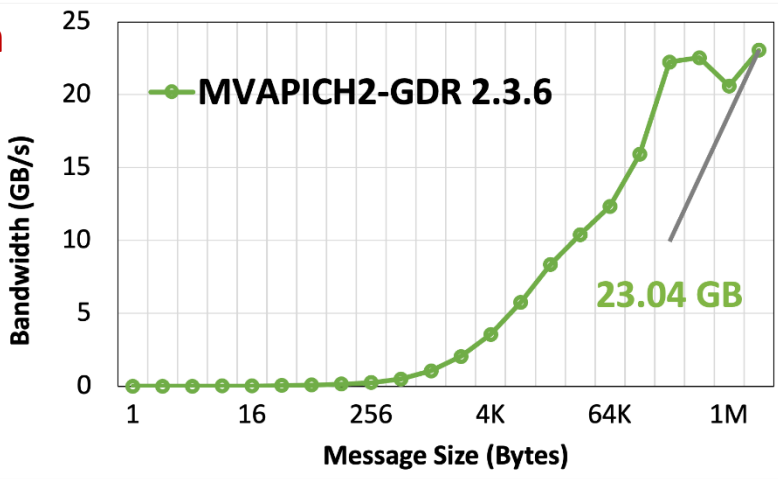


## Inter-Node GPU Point-to-Point

Latency



Bandwidth

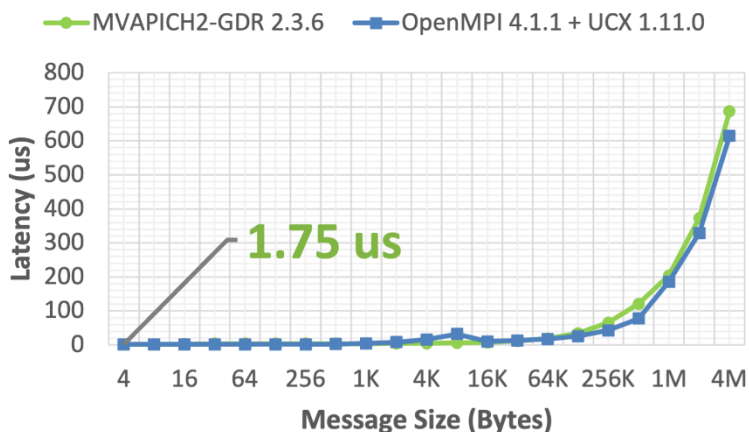


AMD EPYC 7V12 64-Core Processor  
CUDA 11.3, NVIDIA A100 GPUs  
Mellanox ConnectX-6 HDR HCA

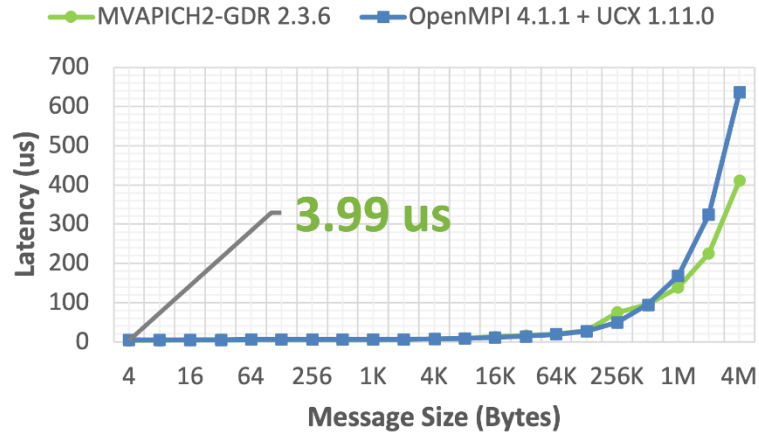


# ROCm-aware MVAPICH2-GDR - Support for AMD GPUs

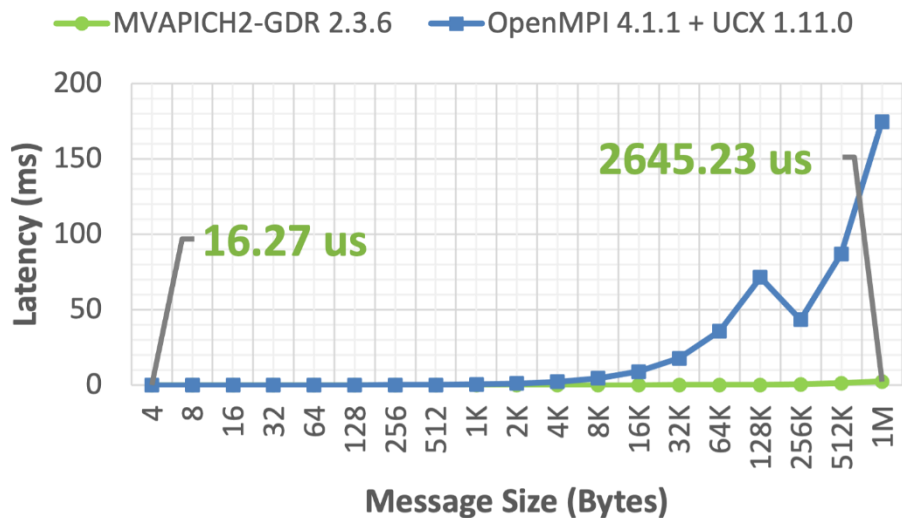
## Intra-Node Point-to-Point Latency



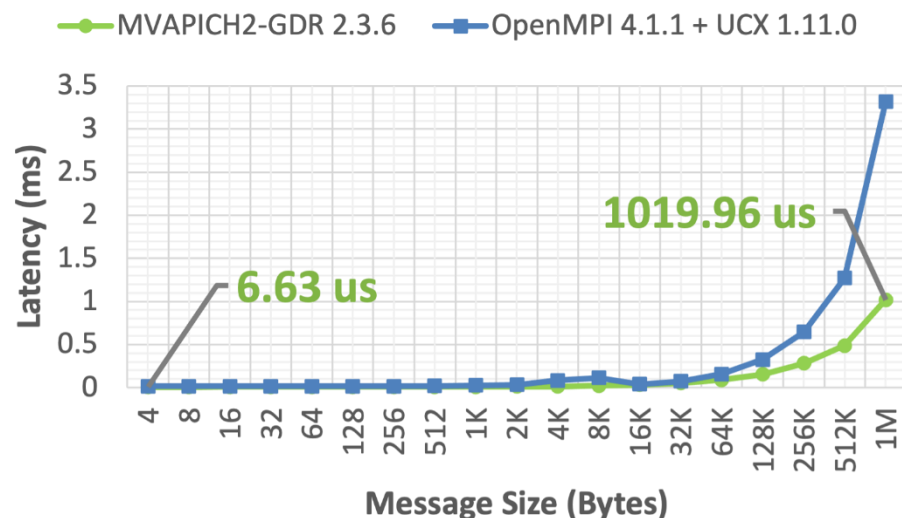
## Inter-Node Point-to-Point Latency



## Allreduce – 64 GPUs (8 nodes, 8 GPUs Per Node)



## Bcast – 64 GPUs (8 nodes, 8 GPUs Per Node)



*More details in the talk "Performance of ROCm-aware MVAPICH2-GDR on LLNL Corona Cluster with AMD GPUs" by Kawthar Shafie Khorassani on Tuesday (08/24/2021) @ 4:30 PM EDT*

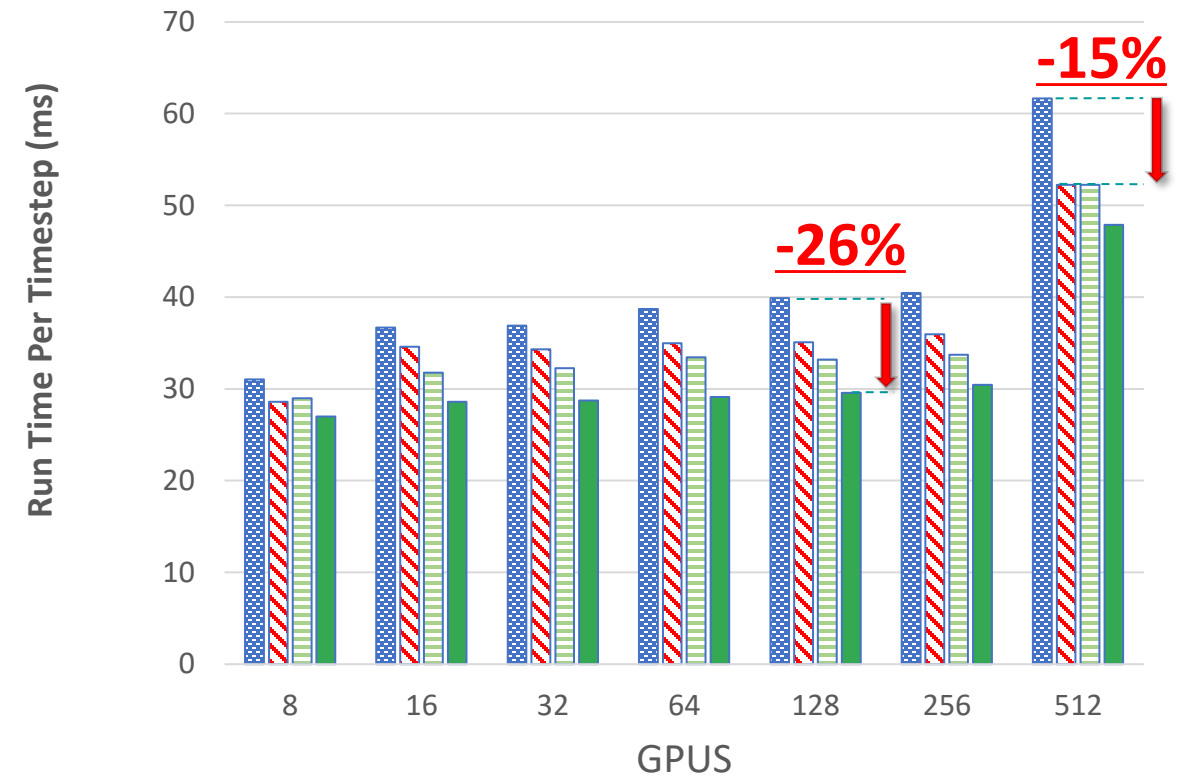
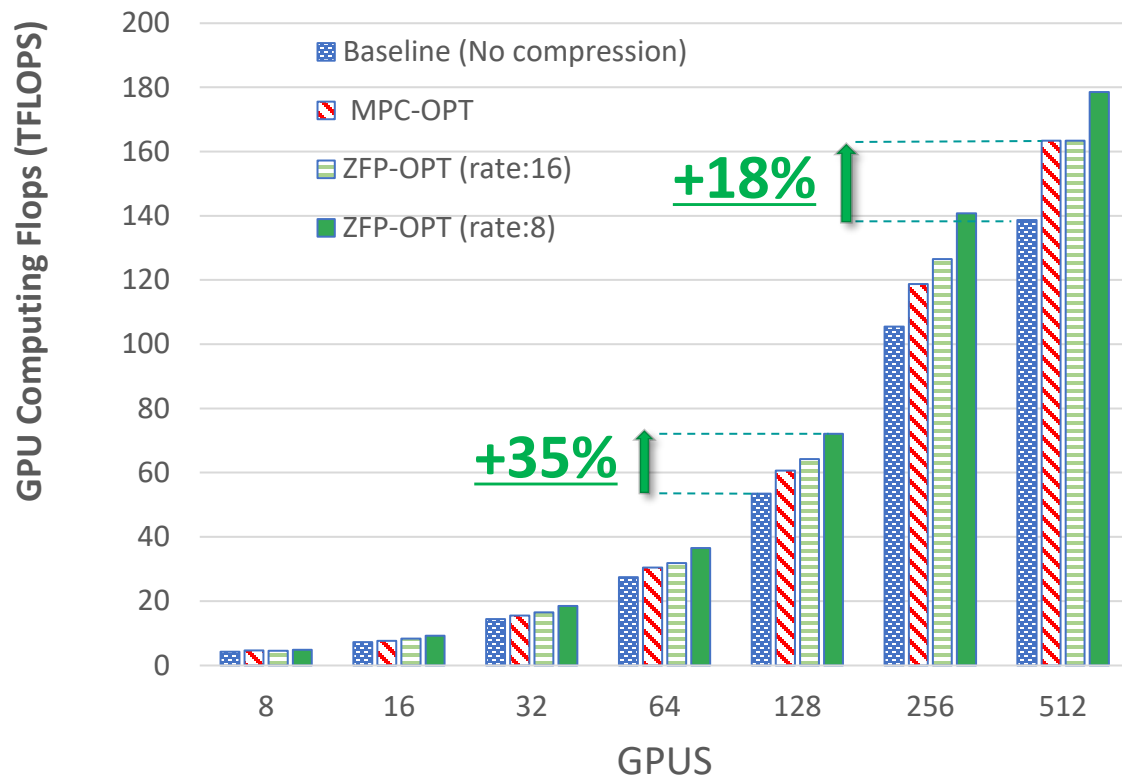
**Corona Cluster @ LLNL - ROCm-4.3.0 (mi50 AMD GPUs)  
Available with MVAPICH2-GDR 2.3.5+ & OMB v5.7+**

K. Khorassani, J. Hashmi, C. Chu, C. Chen, H. Subramoni, D. Panda Designing a ROCm-aware MPI Library for AMD GPUs: Early Experiences - ISC HIGH PERFORMANCE 2021, Jun 2021.

# Performance with “On-the-fly” Compression Support in MVAPICH2-GDR

More details in  
Today's Talk by  
Qinghua Zhou

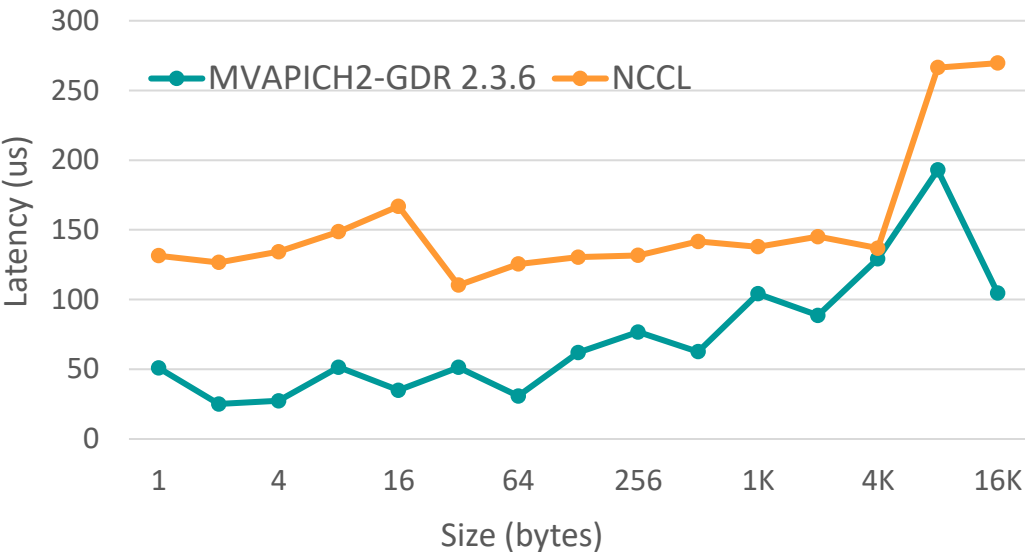
- Weak-Scaling of HPC application **AWP-ODC** on Lassen cluster (V100 nodes) [1]
- MPC-OPT achieves up to **+18%** GPU computing flops, **-15%** runtime per timestep
- ZFP-OPT achieves up to **+35%** GPU computing flops, **-26%** runtime per timestep



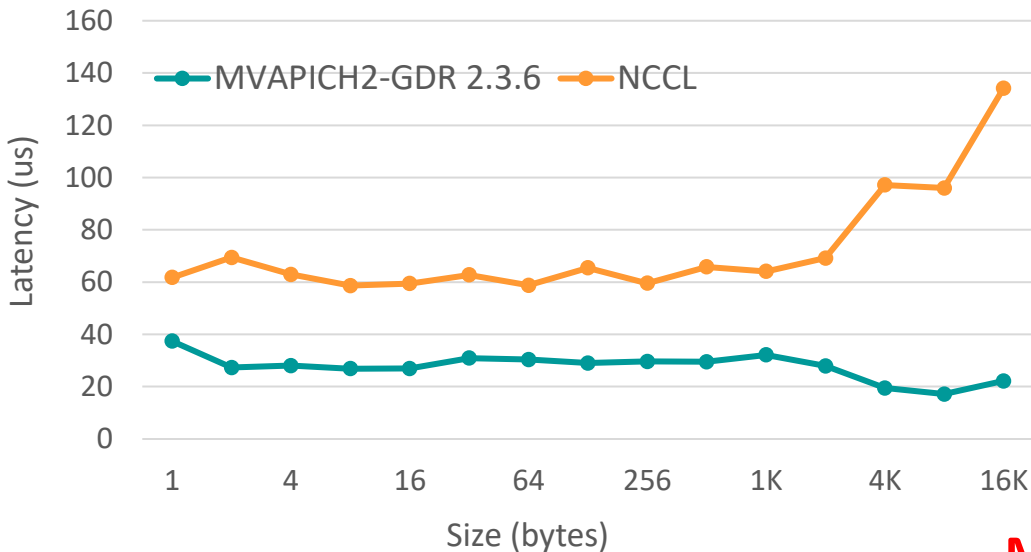
[1] Q. Zhou, C. Chu, N. Senthil Kumar, P. Kousha, M. Ghazimirsaeed, H. Subramoni, and D.K. Panda, "Designing High-Performance MPI Libraries with On-the-fly Compression for Modern GPU Clusters", 35th IEEE International Parallel & Distributed Processing Symposium (IPDPS), May 2021. **[Best Paper Finalist]**

# Collectives Performance on DGX2-A100 – Small Message

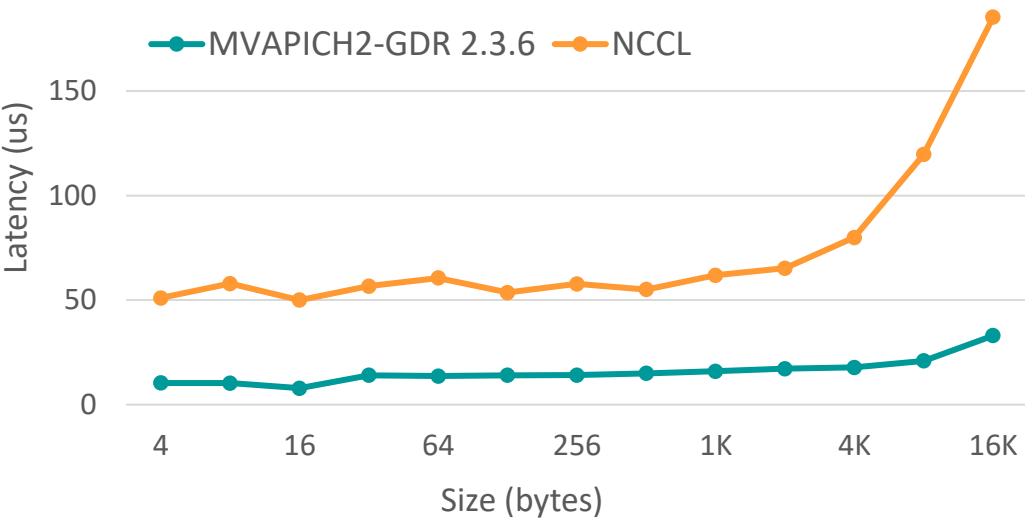
Allgather (4 nodes, 8 ppn, 32 GPUs)



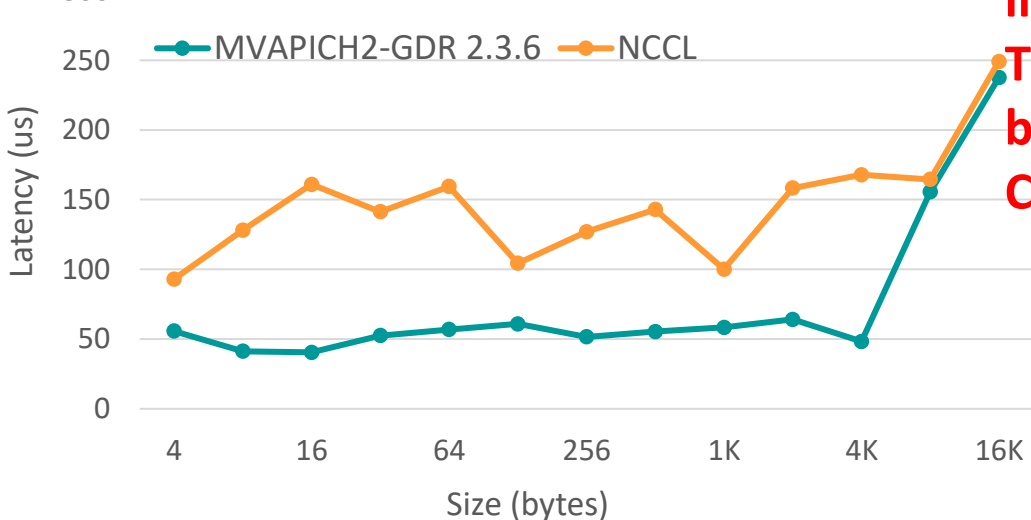
Bcast (4 nodes, 8 ppn, 32 GPUs)



Reduce (4 nodes, 8 ppn, 32 GPUs)

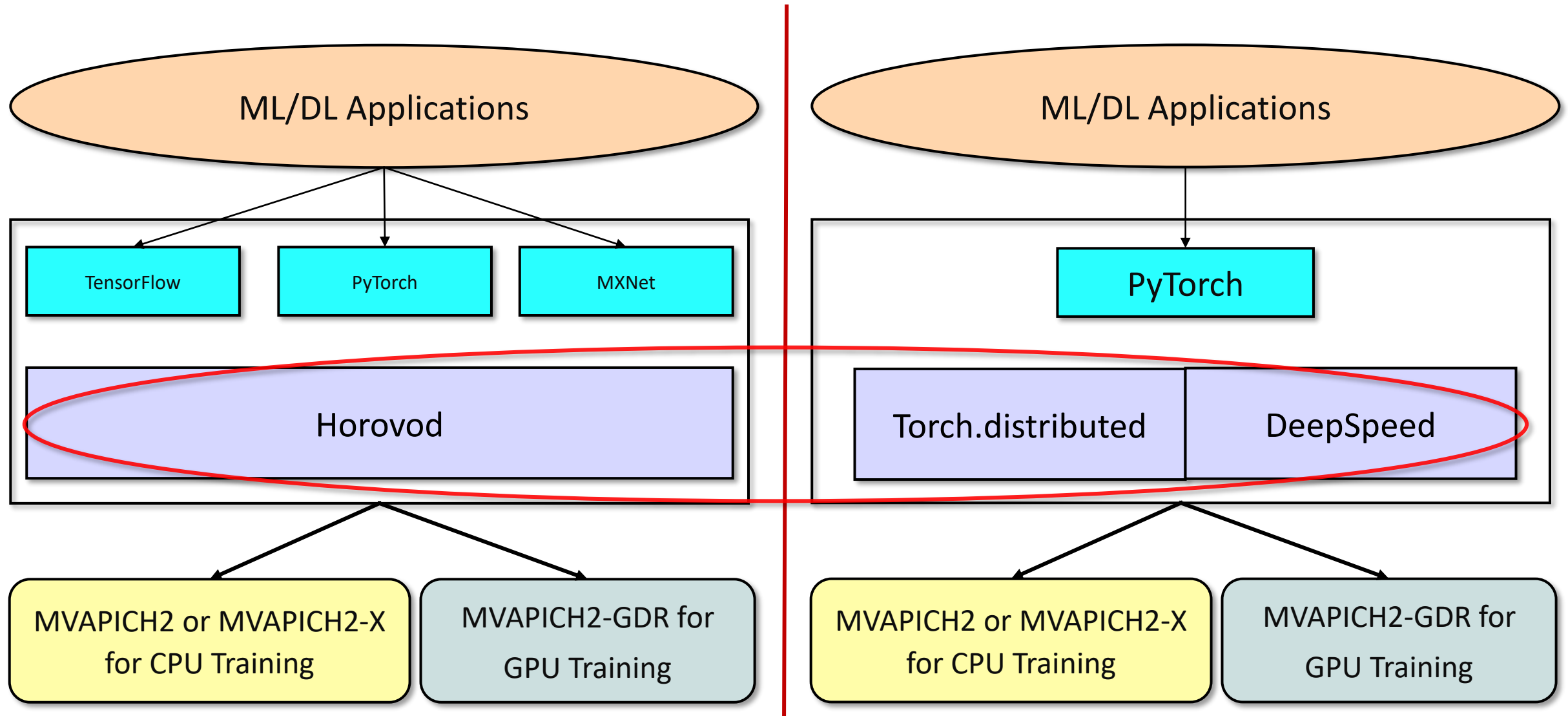


Allreduce (4 nodes, 8 ppn, 32 GPUs)



More details in  
Today's Talk  
by Chen Chun  
Chen

# MVAPICH2 (MPI)-driven Infrastructure for ML/DL Training



More details available from: <http://hidl.cse.ohio-state.edu>

# PyTorch, Horovod and DeepSpeed at Scale: Training ResNet-50 on 256 V100 GPUs

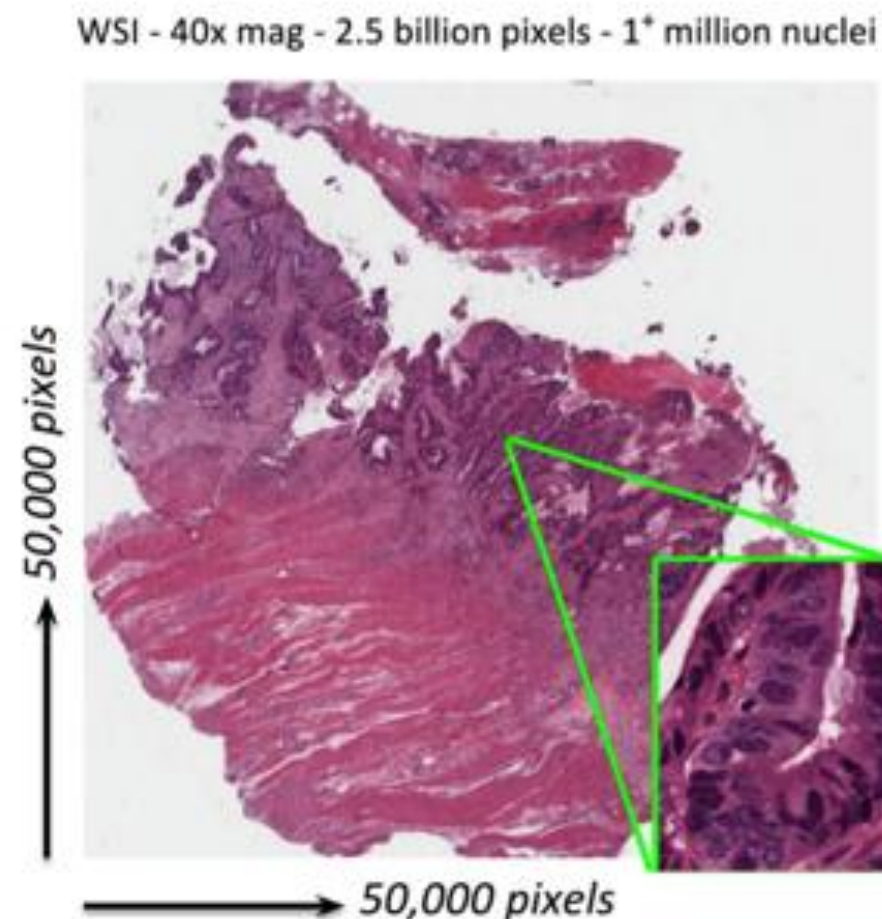
- Training performance for 256 V100 GPUs on LLNL Lassen
  - **~10,000 Images/sec faster** than NCCL training!

| Distributed Framework  | Torch.distributed |               | Horovod |               | DeepSpeed |               |
|------------------------|-------------------|---------------|---------|---------------|-----------|---------------|
| Images/sec on 256 GPUs | 61,794            | <b>72,120</b> | 74,063  | <b>84,659</b> | 80,217    | <b>88,873</b> |
| Communication Backend  | NCCL              | MVAPICH2-GDR  | NCCL    | MVAPICH2-GDR  | NCCL      | MVAPICH2-GDR  |

# Exploiting Model Parallelism in AI-Driven Digital Pathology

- Pathology whole slide image (WSI)
  - Each WSI = 100,000 x 100,000 pixels
  - Can not fit in a single GPU memory
  - Tiles are extracted to make training possible
- Two main problems with tiles
  - Restricted tile size because of GPU memory limitation
  - Smaller tiles loose structural information
- Can we use Model Parallelism to train on larger tiles to get better accuracy and diagnosis?
- **Reduced training time significantly on OpenPOWER + NVIDIA V100 GPUs**
  - **32 hours (1 node, 1 GPU) -> 7.25 hours (1 node, 4 GPUs) -> 27 mins (32 nodes, 128 GPUs)**

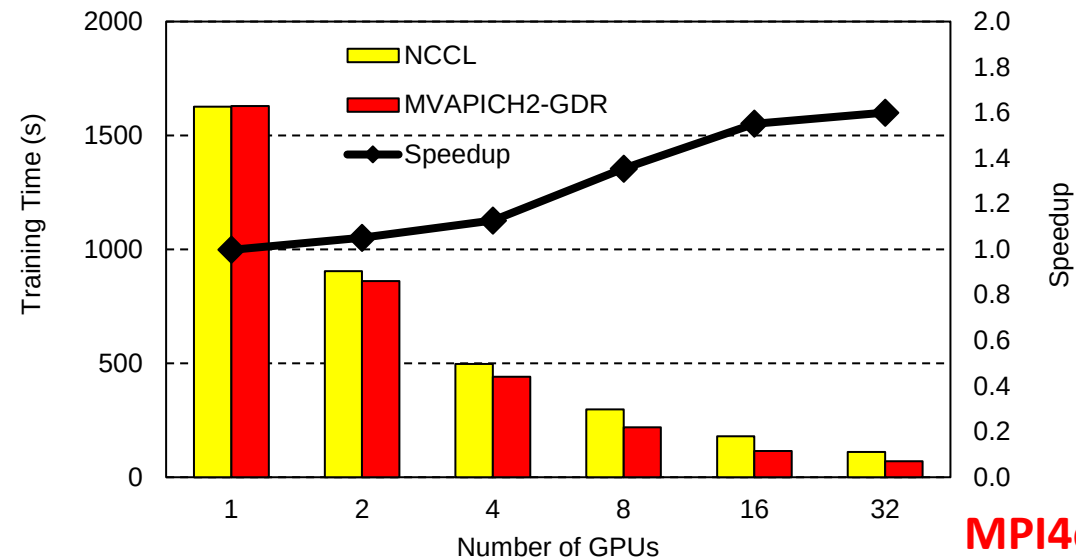
**More details in Tomorrow's talk by Arpan Jain**



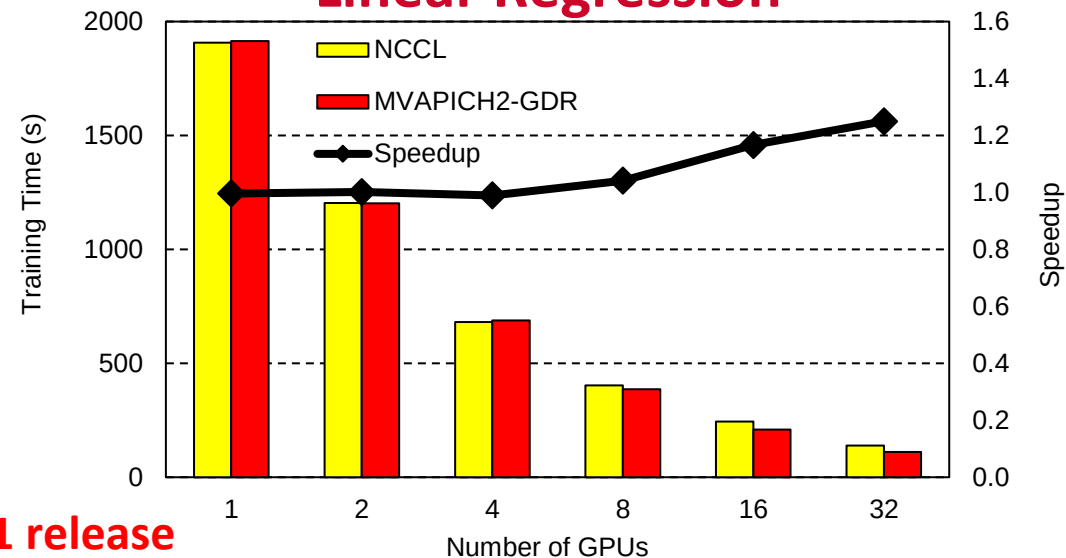
Courtesy: <https://blog.kitware.com/digital-slide-archive-large-image-and-histomicstk-open-source-informatics-tools-for-management-visualization-and-analysis-of-digital-histopathology-data/>

A. Jain, A. Awan, A. Aljuhani, J. Hashmi, Q. Anthony, H. Subramoni, D. K. Panda, R. Machiraju, and A. Parwani, "GEMS: GPU Enabled Memory Aware Model Parallelism System for Distributed DNN Training", Supercomputing (SC '20)

## K-Means

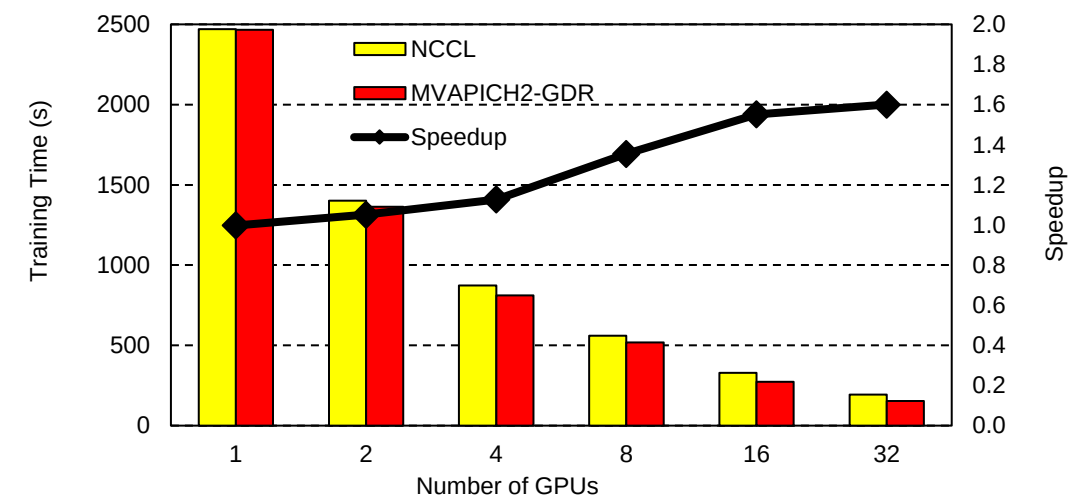


## Linear Regression

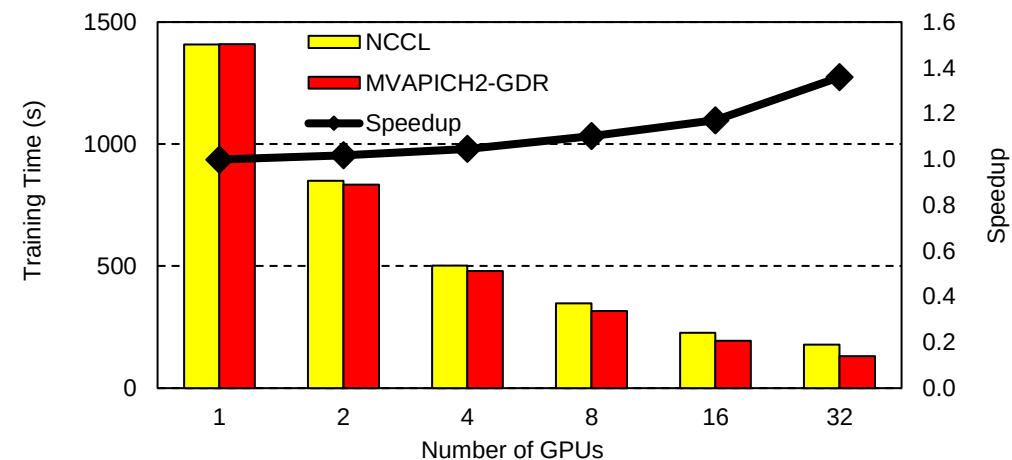


MPI4cuML 0.1 release

## Nearest Neighbors (<http://hidl.cse.ohio-state.edu>)



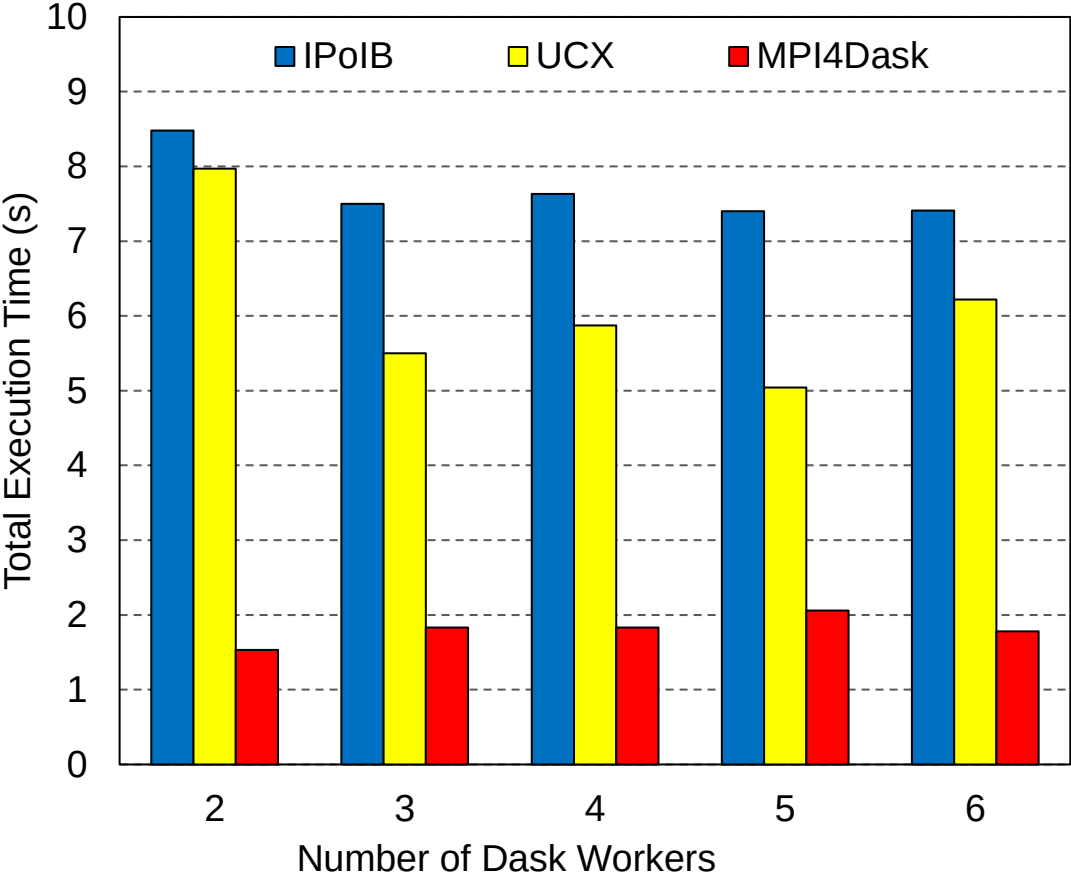
## Truncated SVD



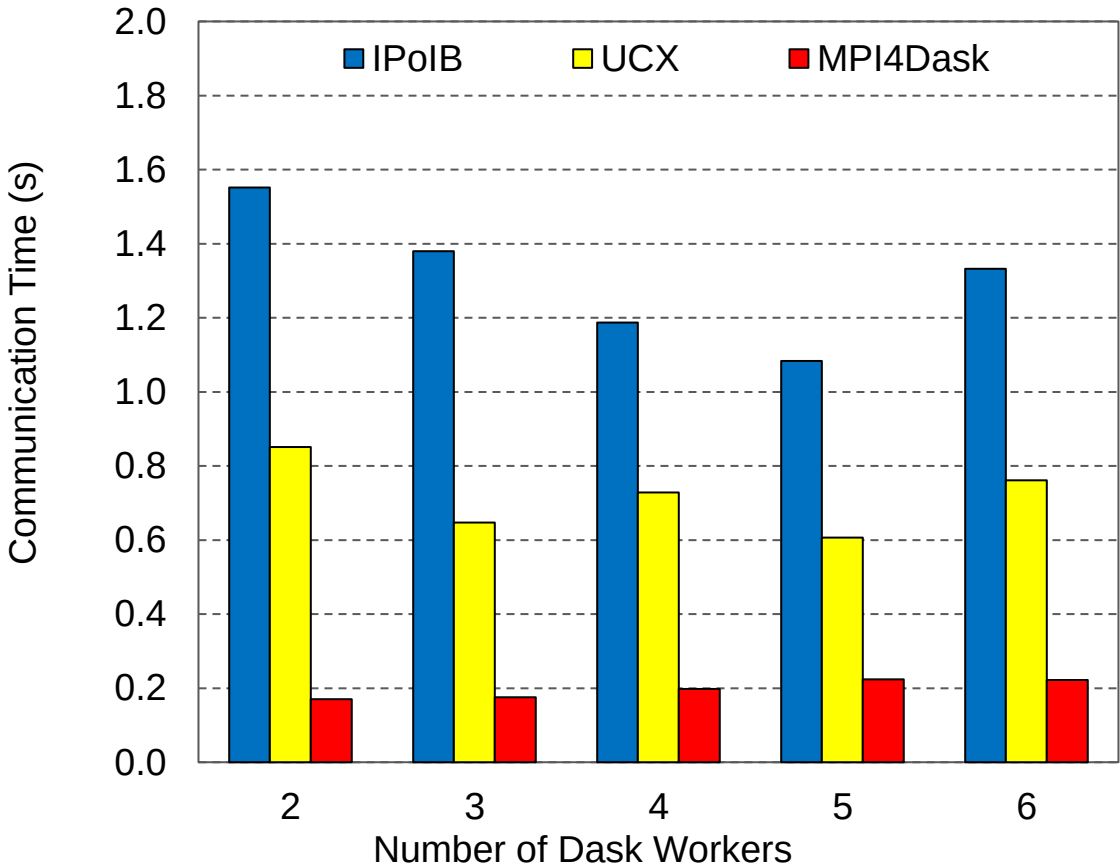
M. Ghazimirsaeed , Q. Anthony , A. Shafi , H. Subramoni , and D. K. Panda, Accelerating GPU-based Machine Learning in Python using MPI Library: A Case Study with MVAPICH2-GDR, MLHPC Workshop, Nov 2020

# Benchmark #1: Sum of cuPy Array and its Transpose (RI2)

3.47x better on average



6.92x better on average



A. Shafi , J. Hashmi , H. Subramoni , and D. K. Panda, Efficient MPI-based Communication for GPU-Accelerated Dask Applications, CCGrid '21, <https://arxiv.org/abs/2101.08878>

MPI4Dask 0.2 release  
(<http://hibd.cse.ohio-state.edu>)



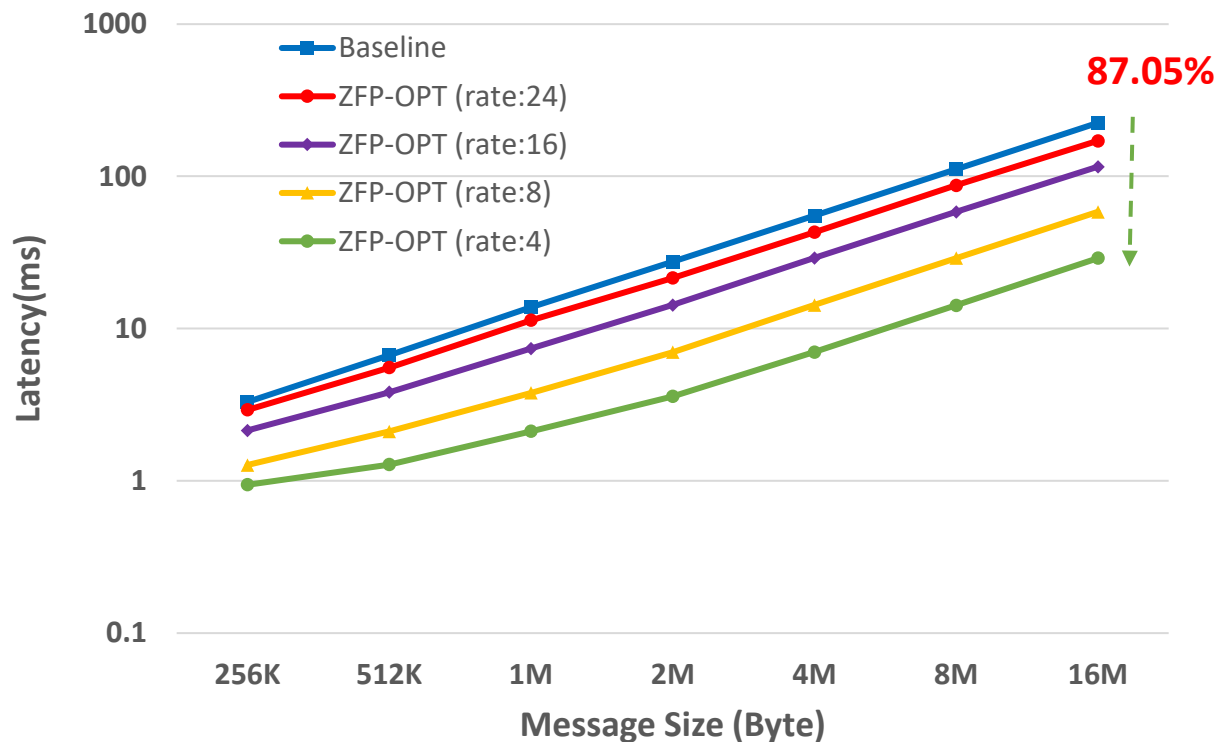
## MVAPICH2-GDR Upcoming Features for HPC and DL

- On-the-fly Compression for All-to-all Collective
- Scalable Distributed Training with Model-/Hybrid Parallelism for out-of-core DNN Models
- Scaling Single-Image Super-Resolution Training

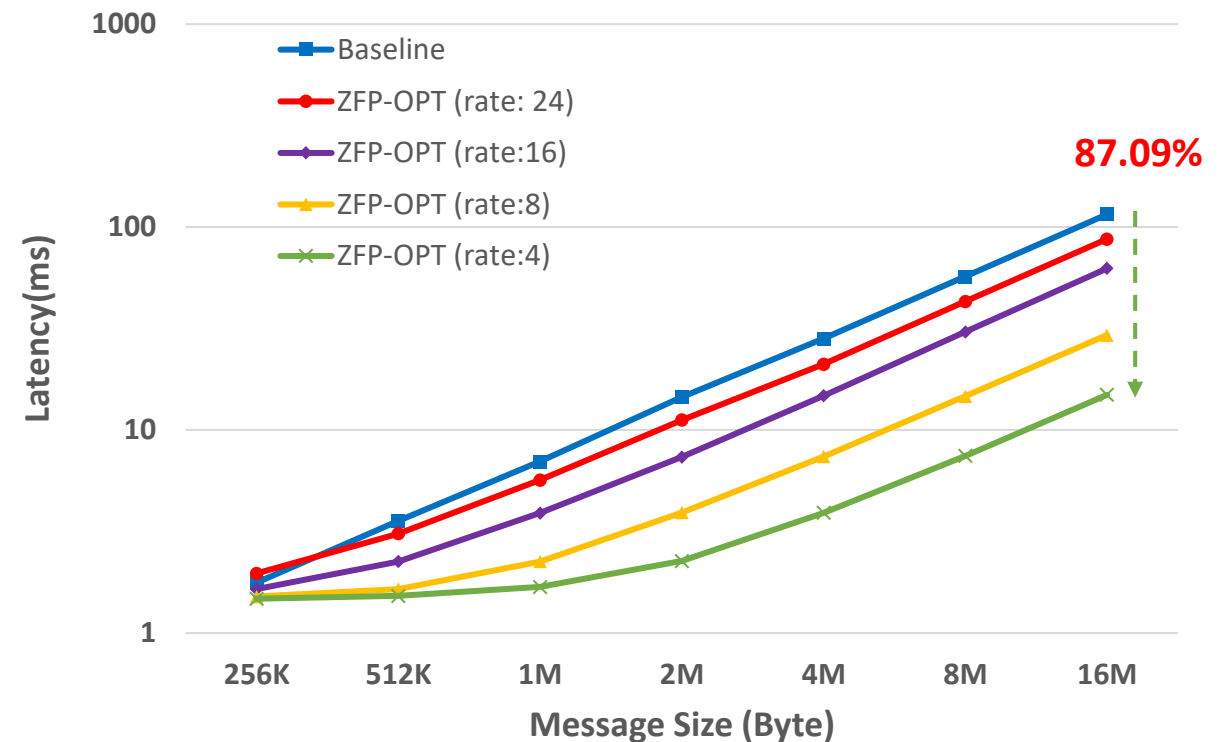
# All-to-All with On-the-Fly Compression

- All-to-All operation with collective level on-the-fly compression design
- ZFP-OPT(rate:4) achieves up to **~87%** reduced latency at 16MB on Frontera RTX-5000 and Longhorn V100 nodes (4nodes, 4ppn)

Frontera-RTX



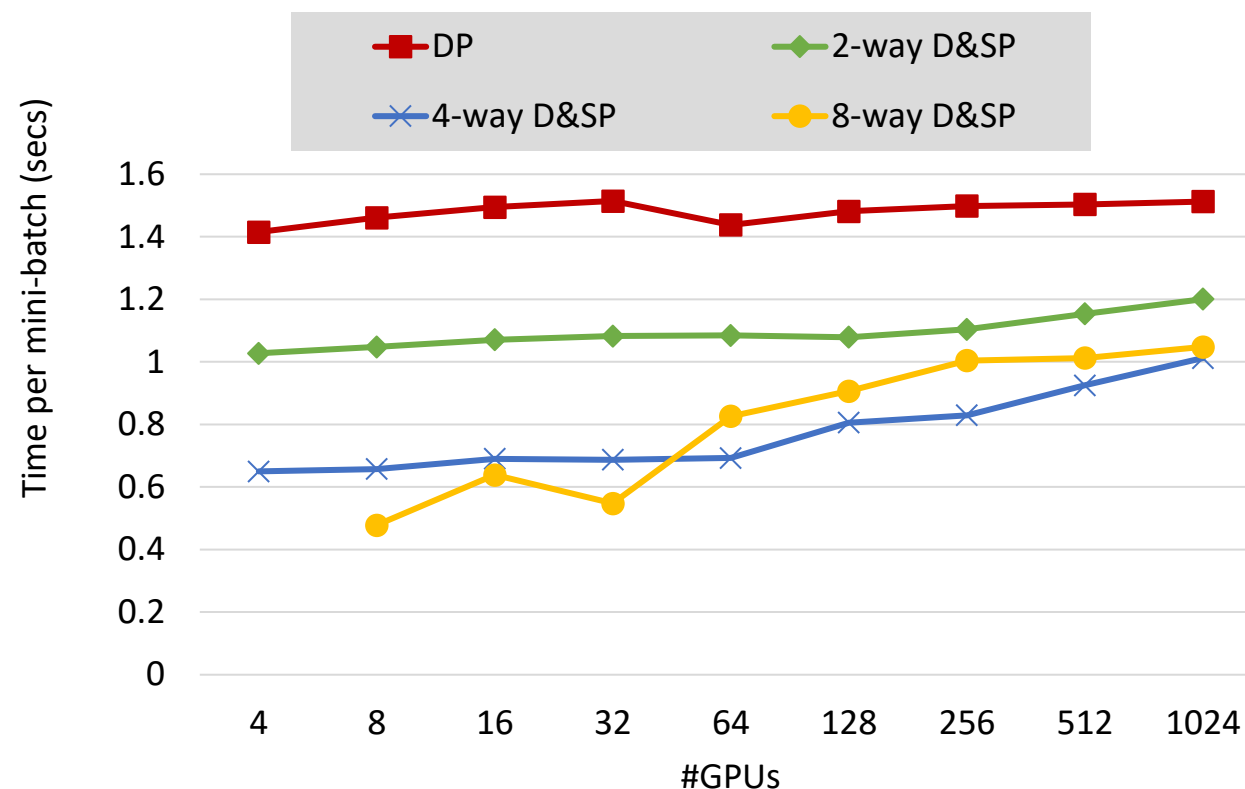
Longhorn-V100



# Accelerating Transformers using SUPER

- We propose sub-graph parallelism integrated with data parallelism to accelerate the training of Transformers.
- Approach
  - Data and Sub-Graph Parallelism (D&SP)
    - #-way D&SP (#: number of sub-graphs)
- Setup
  - T5-Large-Mod on WMT Dataset
  - 1024 NVIDIA V100 GPUs
- Speedup
  - Up to **3.05X** over Data Parallelism (DP)

Up to 3.05X speedup over Data Parallel designs (LLNL Lassen)



More details in Tomorrow's talk by Arpan Jain

# Single-Image Super-Resolution Training: Performance Improvement

More details in Tomorrow's talk  
by Quentin Anthony

- Throughput is improved at all scales
- MPI-Opt performs better than **both** NCCL and default MPI

| #GPUs | Throughput (Img/sec) |      |         | Percentage Improvement<br>(MPI-Opt over default MPI) |
|-------|----------------------|------|---------|------------------------------------------------------|
|       | NCCL                 | MPI  | MPI-Opt |                                                      |
| 1     | 11                   | 11   | 11      | $\approx 0$                                          |
| 2     | 20                   | 21   | 22      | 2.08                                                 |
| 4     | 40                   | 41   | 44      | 6.28                                                 |
| 8     | 76                   | 75   | 83      | 11.15                                                |
| 16    | 142                  | 126  | 156     | 24.26                                                |
| 32    | 269                  | 242  | 296     | 22.32                                                |
| 64    | 511                  | 490  | 588     | 20.05                                                |
| 128   | 989                  | 933  | 1152    | 23.49                                                |
| 256   | 1807                 | 1675 | 2185    | 30.41                                                |
| 512   | 3608                 | 3258 | 4116    | <b>26.33</b>                                         |

Q. Anthony, L. Xu, H. Subramoni, and D. K. Panda, Scaling Single-Image Super-Resolution Training on Modern HPC Clusters: Early Experiences, ScaDL Workshop, in conjunction with IPDPS '21, May 2021

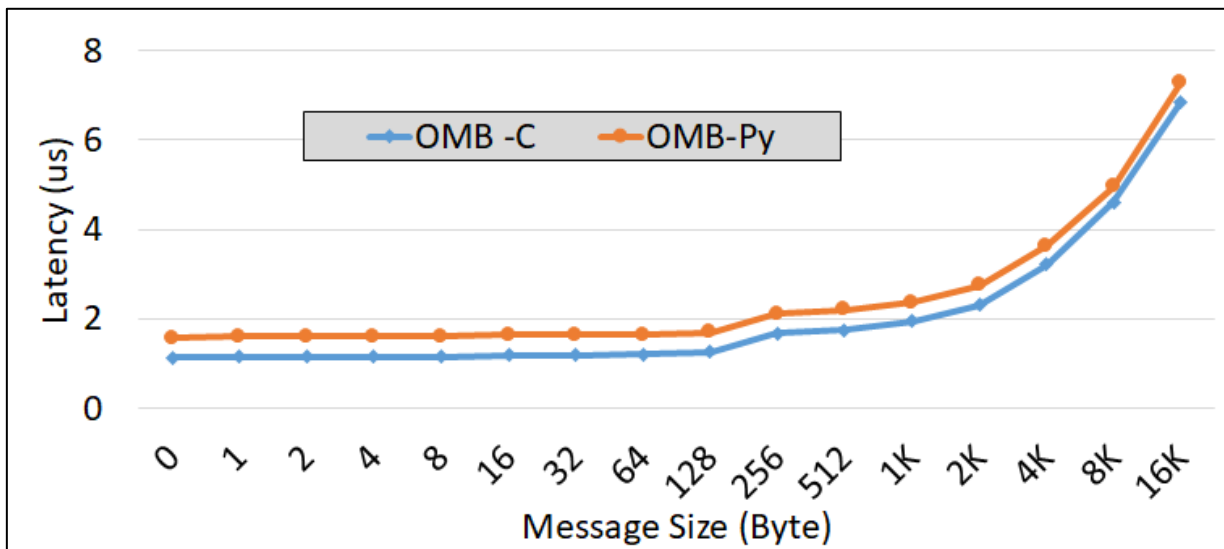
# MVAPICH2 Software Family

| Requirements                                                                                                                   | Library        |
|--------------------------------------------------------------------------------------------------------------------------------|----------------|
| MPI with Support for InfiniBand, Omni-Path, Ethernet/iWARP and, RoCE (v1/v2)                                                   | MVAPICH2       |
| Optimized Support for Microsoft Azure Platform with InfiniBand                                                                 | MVAPICH2-Azure |
| Advanced MPI features/support (UMR, ODP, DC, Core-Direct, SHArP, XPMEM), OSU INAM (InfiniBand Network Monitoring and Analysis) | MVAPICH2-X     |
| Advanced MPI features (SRD and XPMEM) with support for Amazon Elastic Fabric Adapter (EFA)                                     | MVAPICH2-X-AWS |
| Optimized MPI for clusters with NVIDIA GPUs and for GPU-enabled Deep Learning Applications                                     | MVAPICH2-GDR   |
| Energy-aware MPI with Support for InfiniBand, Omni-Path, Ethernet/iWARP and, RoCE (v1/v2)                                      | MVAPICH2-EA    |
| MPI Energy Monitoring Tool                                                                                                     | OEMT           |
| InfiniBand Network Analysis and Monitoring                                                                                     | OSU INAM       |
| Microbenchmarks for Measuring MPI and PGAS Performance                                                                         | OMB            |

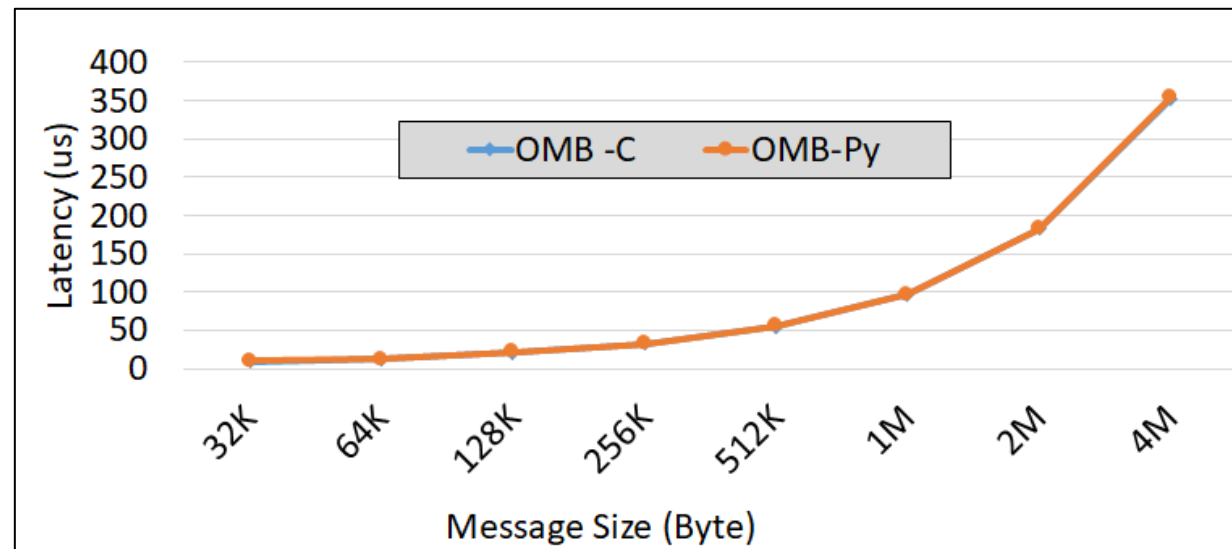
# OSU Microbenchmarks

- Available since 2004
- Suite of microbenchmarks to study communication performance of various programming models
- Benchmarks available for the following programming models
  - Message Passing Interface (MPI)
  - Partitioned Global Address Space (PGAS)
    - Unified Parallel C (UPC), Unified Parallel C++ (UPC++), and OpenSHMEM
- Benchmarks available for multiple accelerator-based architectures
  - Compute Unified Device Architecture (CUDA)
  - OpenACC Application Program Interface
- Part of various national resource procurement suites like NERSC-8 / Trinity Benchmarks
- Continuing to add support for newer primitives and features (latest OMB 5.8)
  - ROCm support
  - NCCL support
- Will be adding Python and Java MPI support in OMB
- Please visit the following link for more information: <http://mvapich.cse.ohio-state.edu/benchmarks/>

# OMB-Py: Initial CPU Results



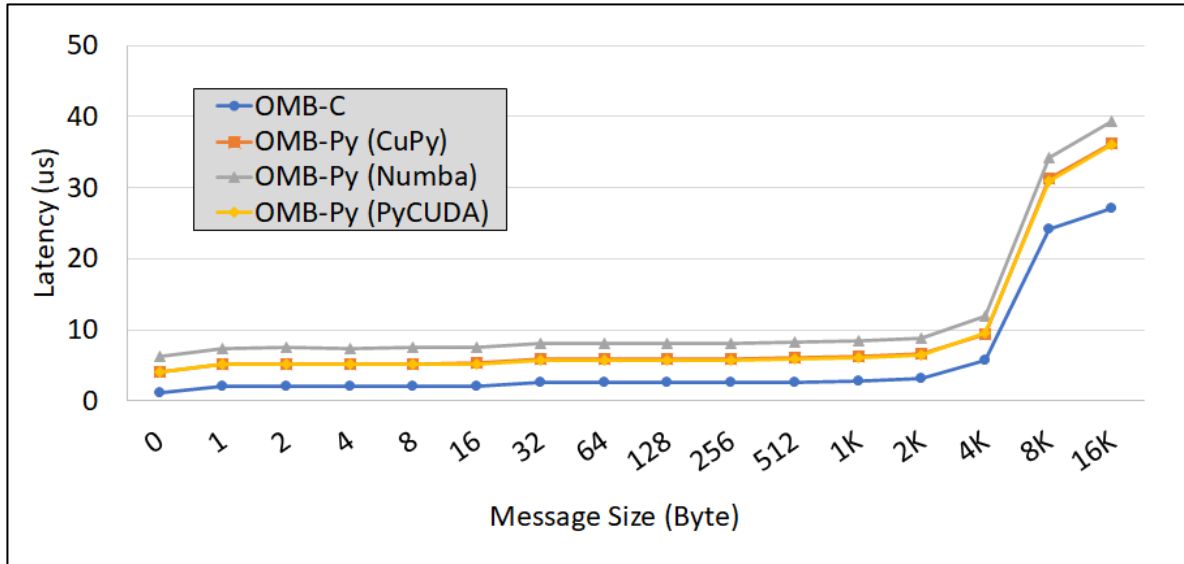
osu\_latency inter-node small messages



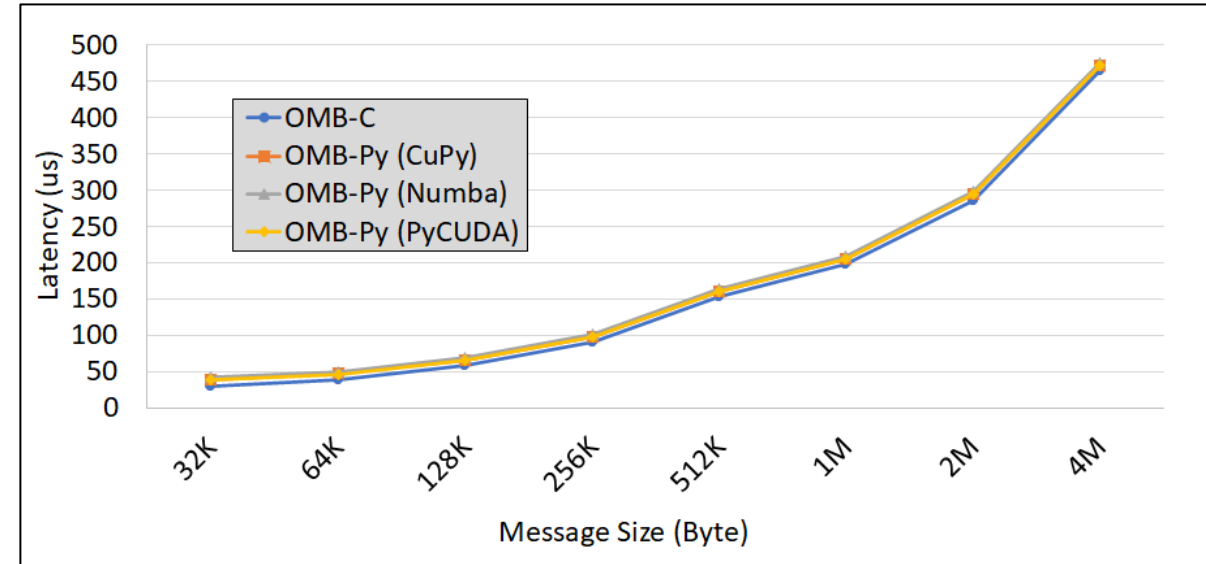
osu\_latency inter-node large messages

- Small Python overhead around 0.5 microseconds.
- Overhead only apparent in smaller message sizes

# OMB-Py: Initial GPU Results



osu\_latency inter-node small messages



osu\_latency inter-node large messages

- Different overheads depending on the CUDA-aware data structure
- Overhead is only apparent in smaller message sizes

**To be available with the  
next OMB release**

**More details in  
Tomorrow's Talk  
by Nawras Alnaasan**



# Applications-Level Tuning: Compilation of Best Practices

- MPI runtime has many parameters
- Tuning a set of parameters can help you to extract higher performance
- Compiled a list of such contributions through the MVAPICH Website
  - [http://mvapich.cse.ohio-state.edu/best\\_practices/](http://mvapich.cse.ohio-state.edu/best_practices/)
- Initial list of applications
  - Amber
  - HoomDBlue
  - HPCG
  - Lulesh
  - MILC
  - Neuron
  - SMG2000
  - Cloverleaf
  - SPEC (LAMMPS, POP2, TERA\_TF, WRF2)
- Soliciting additional contributions, send your results to mvapich-help at cse.ohio-state.edu.
- We will link these results with credits to you.

# MVAPICH2 Libraries available through Spack

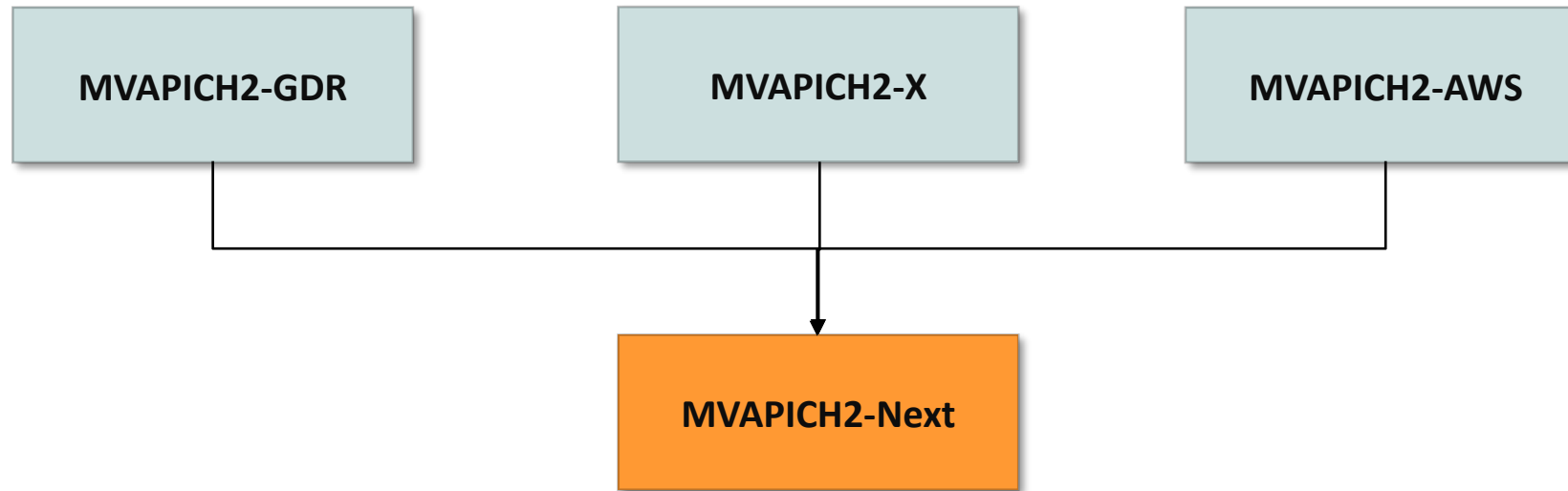
- Easy Installation of MVAPICH2 Libraries through Spack
  - MVAPICH2
  - MVAPICH2-X
  - MVAPICH2-GDR
- Detailed Spack-based Installation User Guide is available:  
[http://mvapich.cse.ohio-state.edu/userguide/userguide\\_spack/](http://mvapich.cse.ohio-state.edu/userguide/userguide_spack/)

**More details in  
Tomorrow's Talk  
by Todd Gamblin (LLNL)**

## Enabling Auto detection of Underlying features with dlopen

- Use dlopen to identify support available in underlying system
  - Removes dependencies on libraries like
    - ibverbs
    - ibmad
    - ibumad
    - rdmacm
    - xpmem
  - Simplifies installation and deployment
    - No need for multiple configure options
    - One binary can have all necessary support
- Already available with the latest MVAPICH2 releases
- More such functionalities will be added in future releases

# MVAPICH2-Next: One Stack to Conquer all Architectures and Interconnects



- Various libraries will be merged into one release
- Simplifies installation and deployment
- Enables utilizing the best advanced features for all architectures

# MVAPICH2 – Plans for Exascale

- Performance and Memory scalability toward 1-10M cores
- Hybrid programming (MPI + OpenSHMEM, MPI + UPC, MPI + CAF ...)
  - MPI + Task\*
- Enhanced Optimization for GPU Support and Accelerators
- Taking advantage of advanced features of Mellanox InfiniBand
  - Tag Matching\*
  - Adapter Memory\*
- Enhanced communication schemes for upcoming architectures
  - Intel Optane\*
  - CAPI\*
- Extended topology-aware collectives
- Extended Energy-aware designs and Virtualization Support
- Extended Support for MPI Tools Interface (as in MPI 3.0)
- Extended FT support
- Support for \* features will be available in future MVAPICH2 Releases

# Commercial Support for MVAPICH2, HiBD, and HiDL Libraries

- Supported through X-ScaleSolutions (<http://x-scalesolutions.com>)
- Benefits:
  - Help and guidance with installation of the library
  - Platform-specific optimizations and tuning
  - Timely support for operational issues encountered with the library
  - Web portal interface to submit issues and tracking their progress
  - Advanced debugging techniques
  - Application-specific optimizations and tuning
  - Obtaining guidelines on best practices
  - Periodic information on major fixes and updates
  - Information on major releases
  - Help with upgrading to the latest release
  - Flexible Service Level Agreements
- Support being provided to National Laboratories and International HPC centers



# Value-Added Products with Support

- Multiple value-added products with support
  - X-ScaleHPC
  - X-ScaleAI
  - MVAPICH2-DPU (Tutorial and Live Demo yesterday)
  - SCR-Exa

More details in Donglai Dai's Talk (Tomorrow)



# Funding Acknowledgments

## Funding Support by



## Equipment Support by





# Acknowledgments to all the Heroes (Past/Current Students and Staffs)

## ***Current Students (Graduate)***

- |                       |                            |                         |
|-----------------------|----------------------------|-------------------------|
| – N. Alnaasan (Ph.D.) | – K. S. Khorassani (Ph.D.) | – A. H. Tu (Ph.D.)      |
| – Q. Anthony (Ph.D.)  | – P. Kousha (Ph.D.)        | – S. Xu (Ph.D.)         |
| – C.-C. Chun (Ph.D.)  | – B. Michalowicz (Ph.D.)   | – Q. Zhou (Ph.D.)       |
| – N. Contini (Ph.D.)  | – B. Ramesh (Ph.D.)        | – K. Al Attar (M.S.)    |
| – A. Jain (Ph.D.)     | – K. K. Suresh (Ph.D.)     | – N. Sarkauskas (Ph.D.) |

## ***Current Research Scientists***

- A. Shafi
- H. Subramoni

## ***Current Software Engineers***

- B. Seeds
- N. Shineman

## ***Current Students (Undergrads)***

- M. Lieber
- L. Xu

## ***Past Students***

- |                             |                            |                               |
|-----------------------------|----------------------------|-------------------------------|
| – A. Awan (Ph.D.)           | – T. Gangadharappa (M.S.)  | – P. Lai (M.S.)               |
| – A. Augustine (M.S.)       | – K. Gopalakrishnan (M.S.) | – J. Liu (Ph.D.)              |
| – P. Balaji (Ph.D.)         | – J. Hashmi (Ph.D.)        | – M. Luo (Ph.D.)              |
| – M. Bayatpour (Ph.D.)      | – W. Huang (Ph.D.)         | – A. Mamidala (Ph.D.)         |
| – R. Biswas (M.S.)          | – W. Jiang (M.S.)          | – G. Marsh (M.S.)             |
| – S. Bhagvat (M.S.)         | – J. Jose (Ph.D.)          | – V. Meshram (M.S.)           |
| – A. Bhat (M.S.)            | – M. Kedia (M.S.)          | – A. Moody (M.S.)             |
| – D. Buntinas (Ph.D.)       | – S. Kini (M.S.)           | – S. Naravula (Ph.D.)         |
| – L. Chai (Ph.D.)           | – M. Koop (Ph.D.)          | – R. Noronha (Ph.D.)          |
| – B. Chandrasekharan (M.S.) | – K. Kulkarni (M.S.)       | – X. Ouyang (Ph.D.)           |
| – S. Chakraborty (Ph.D.)    | – R. Kumar (M.S.)          | – S. Pai (M.S.)               |
| – N. Dandapanthula (M.S.)   | – S. Krishnamoorthy (M.S.) | – S. Potluri (Ph.D.)          |
| – V. Dhanraj (M.S.)         | – K. Kandalla (Ph.D.)      | – K. Raj (M.S.)               |
| – C.-H. Chu (Ph.D.)         | – M. Li (Ph.D.)            | – R. Rajachandrasekar (Ph.D.) |

- D. Shankar (Ph.D.)
- G. Santhanaraman (Ph.D.)
- N. Sarkauskas (B.S.)
- N. Senthil Kumar (M.S.)
- A. Singh (Ph.D.)
- J. Sridhar (M.S.)
- S. Srivastava (M.S.)
- S. Sur (Ph.D.)
- H. Subramoni (Ph.D.)
- K. Vaidyanathan (Ph.D.)
- A. Vishnu (Ph.D.)
- J. Wu (Ph.D.)
- W. Yu (Ph.D.)
- J. Zhang (Ph.D.)

## ***Past Research Scientists***

- K. Hamidouche
- S. Sur
- X. Lu

## ***Past Senior Research Associate***

- J. Hashmi

## ***Past Programmers***

- A. Reifsteck
- D. Bureddy
- J. Perkins

## ***Past Research Specialist***

- M. Arnold
- J. Smith

## ***Past Post-Docs***

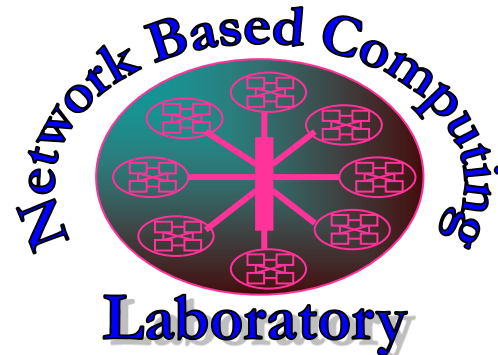
- |                        |             |                 |             |
|------------------------|-------------|-----------------|-------------|
| – D. Banerjee          | – H.-W. Jin | – E. Mancini    | – A. Ruhela |
| – X. Besseron          | – J. Lin    | – K. Manian     | – J. Vienne |
| – M. S. Ghazimeersaeed | – M. Luo    | – S. Marcarelli | – H. Wang   |

## Multiple Positions Available in MVAPICH2, BigData and DL/ML Projects in my Group

- Looking for Bright and Enthusiastic Personnel to join as
  - PhD Students
  - Post-Doctoral Researchers
  - MPI Programmer/Software Engineer
  - Spark/Big Data Programmer/Software Engineer
  - Deep Learning, Machine Learning, and Cloud Programmer/Software Engineer
- If interested, please send an e-mail to [panda@cse.ohio-state.edu](mailto:panda@cse.ohio-state.edu)

# Thank You!

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Network-Based Computing Laboratory

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



The High-Performance MPI/PGAS Project

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

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High-Performance  
Big Data

The High-Performance Big Data Project

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



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