

# A Peak Performance Model for All-to-all on Hierarchical Systems and Its Applications

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# Introduction

- All-to-all (A2A) communication: crucial component of many scientific computing applications

- ✓ Fluid dynamics: Direct Numerical Simulations (3D FFTs with distributed transposes)
- ✓ Machine learning
- ✓ ...

- Increasingly massive parallelism  $\implies$  higher fidelity, but A2A renders the application network bandwidth bound.

Scalability challenge  $\iff$  A2A performance

## “Peak” performance model

A theoretical upper bound for all-to-all performance on a given system

Basis: **hierarchical structure** of emergent heterogeneous HPC platforms.

Key feature: model definition and parameters based on easily accessible **system/network specifications**.

# Hierarchical Communication Pathways



“self”

High-bandwidth memory



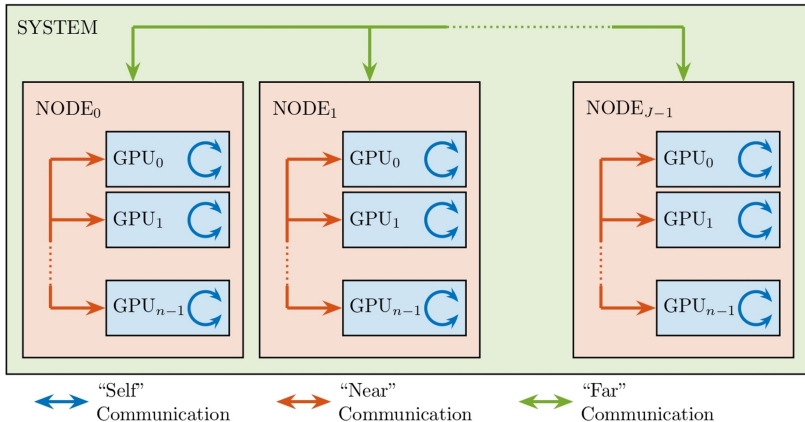
“near”

NVLink, Infinity Fabric, etc.



“far”

Infiniband, Slingshot, etc.



# All-to-All Performance Model

- $P$  processes globally, tightly connected groups of  $n$  processes, A2A message buffer of size  $m$ , and p2p messages of size  $m/P$ .

	“self”	“near”	“far”
# p2p Messages	1	$(n - 1)$	$(P - n)$
Bandwidth	$B_{\text{self}}$	$B_{\text{near}}$	$B_{\text{far}}$
A2A Time	$t_{\text{self}} = \frac{m}{P} \frac{1}{B_{\text{self}}}$	$t_{\text{near}} = (n - 1) \frac{m}{P} \frac{1}{B_{\text{near}}}$	$t_{\text{far}} = (P - n) \frac{m}{P} \frac{1}{B_{\text{far}}}$

- Assumption: concurrent p2p communication along all 3 pathways.

## Peak All-to-All Performance Model

$$\text{Time: } t_{\text{a2a}} = \max \left\{ (P - n) \frac{m}{P} \frac{1}{B_{\text{far}}}, (n - 1) \frac{m}{P} \frac{1}{B_{\text{near}}}, \frac{m}{P} \frac{1}{B_{\text{self}}} \right\}$$

$$\text{Peak Bandwidth: } B_{\text{a2a}} = m/t_{\text{a2a}}$$

# Considerations while Applying the Model

## Peak All-to-All Performance Model

$$\text{Time: } t_{a2a} = \max \left\{ (P - n) \frac{m}{P} \frac{1}{B_{\text{far}}}, (n - 1) \frac{m}{P} \frac{1}{B_{\text{near}}}, \frac{m}{P} \frac{1}{B_{\text{self}}} \right\}$$

$$\text{Peak Bandwidth: } B_{a2a} = m/t_{a2a}$$

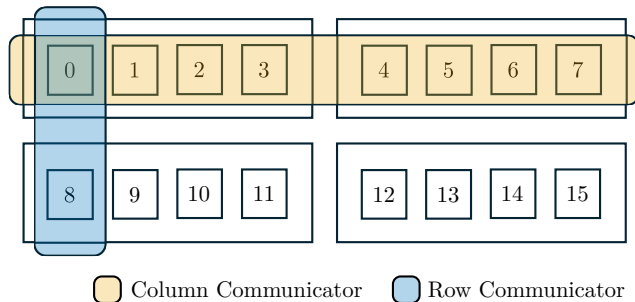
- 1 “near” group definition **dependent on application and node topology.**
- 2 Model parameters: **number of p2p messages and bandwidths** of each communication pathway based on **system specifications.**

# Application: 2D Distributed Transposes

- Data distributed over  $P$  processes, decomposed into  $P_r$  rows and  $P_c$  columns.
- A2A over row and column communicators of different sizes and configurations.
- **Key caveat:** communicators may be split between different nodes, affects definition of “near” term.

E.g.,  $P = P_r \times P_c = 2 \times 8$ , on a system with 4 GPUs per node, A2A time:

$$t_{a2a}^{2D} = t_{col} + t_{row} = t_{a2a}(P = 8, n = 4) + t_{a2a}(P = 2, n = 1)$$



# Test Systems: *Vista*, *Alps* & NVL72

	<i>Vista</i>	<i>Alps</i>	NVL72
GPUs/node	1 NVIDIA GH200	4 NVIDIA GH200	4 NVIDIA GB200
“Far” BW (per GPU)	InfiniBand 50 GB/s	Slingshot 25 GB/s	InfiniBand 50 GB/s
“Near” BW (GPU-GPU)	–	NVLink 150 GB/s	NVLink and NVSwitch 900 GB/s
GPUs/“Near” group	–	4	72
“Self” BW per GPU	HBM3 4 TB/s		HBM3e 8 TB/s

“near” group size on *Vista*  $n = 1 \implies t_{\text{near}} = 0$ .

# Applying the A2A Model on *Alps*

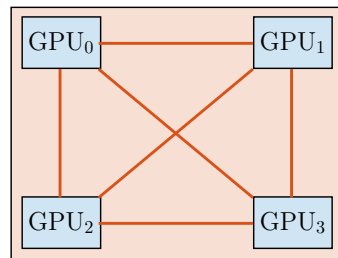
	GPUs/node	"Far" BW (per GPU)	"Near" BW (GPU-GPU)	GPUs/"Near" group	"Self" BW per GPU
<i>Alps</i>	4 NVIDIA GH200	Slingshot 25 GB/s	NVLink 150 GB/s	4	HBM3 4 TB/s

- On a node, each NVLink connected GPU-GPU pair can communicate over a 150 GB/s link.
- But the GPUs split their NVLink connections across peers: max. NVLink BW achieved only when communicating with all 3 peers.

$\Rightarrow B_{\text{near}}$  is a function of the number of "near" processes participating in A2A:

$$B_{\text{near}}(n) = (n - 1) \times 150 \text{ GB/s}$$

**Key takeaway:** Choice of  $B_{\text{near}}$  must account for the node topology.



Node topology on *Alps*



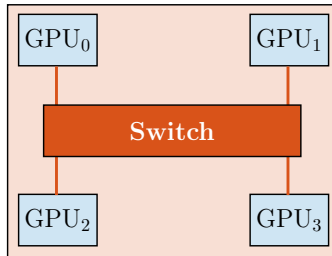
# Applying the A2A Model on the NVL72 System

	GPUs/node	"Far" BW (per GPU)	"Near" BW (GPU-GPU)	GPUs/"Near" group	"Self" BW per GPU
NVL72	4 NVIDIA GB200	InfiniBand 50 GB/s	NVLink and NVSwitch 900 GB/s	72	HBM3e 8 TB/s

- Interconnected NVL72 domains, each consisting of 18 nodes with 4 GPUs each.
- Each domain  $\equiv$  single, large, 72 GPU node.

$\Rightarrow$  "near" pathway is *intra-domain*,  
"far" pathway is *inter-domain*.

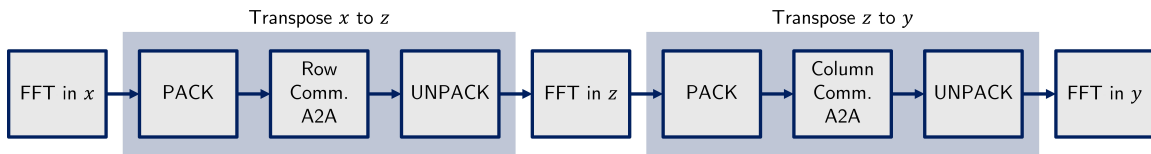
$$B_{\text{near}} = 900 \text{ GB/s}, \quad B_{\text{far}} = 50 \text{ GB/s}$$



Node topology on NVL72

# Benchmark Code: GPU-enabled Direct Numerical Simulations

- Pseudospectral algorithm for direct numerical simulations of turbulent fluid flows.
- GPU algorithm: Yeung et al., *Computer Physics Communications*, 2025 (Fortran, OpenMP offloading, GPU-aware MPI).
- A2A communication required for **distributed transposes** (1D or 2D) between 1D FFTs in three coordinate directions.
- 3D solution domain with  $N^3$  grid points distributed among  $P$  processes, A2A message size  $m = 4N^3/P$ .
- A2A communication dominates time/step ( $\geq 80\%$ ).



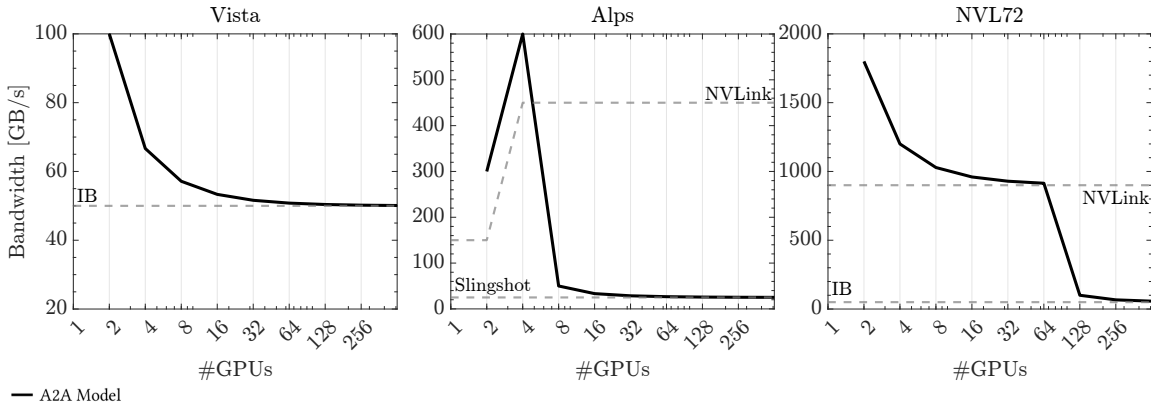
# Testing Multiple A2A Backends with cuDecomp

- Adaptive 2D Decomposition library (Romero *et al.* PASC 2023).
- Multiple A2A backends for global transposition, to compare with model prediction.
- Key performance data collected: average runtime and bandwidth of A2A communication and local pack/unpack in distributed transpose.

Backend	Communication APIs
MPI_P2P MPI_P2P (pipelined)	MPI_Isend, MPI_Irecv
MPI_A2A	MPI_Alltoall, MPI_Alltoallv
NCCL NCCL (pipelined)	ncclSend, ncclRecv
NVSHMEM NVSHMEM (pipelined)	nvshmemx_putmem_nbi_on_stream, nvshmemx_putmem_nbi

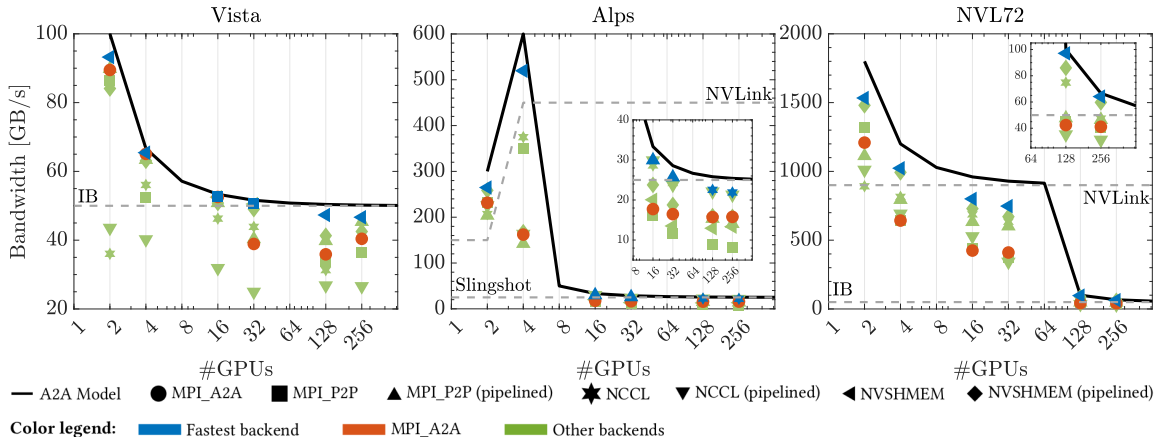
# All-to-all Model Prediction

- Gradual transitions between “self”, “near” and “far” communication dominated regions.
- Performance boosts from “self” and “near” communication at all scales.
- Transition regions pushed to larger scales as “near” group size increases.



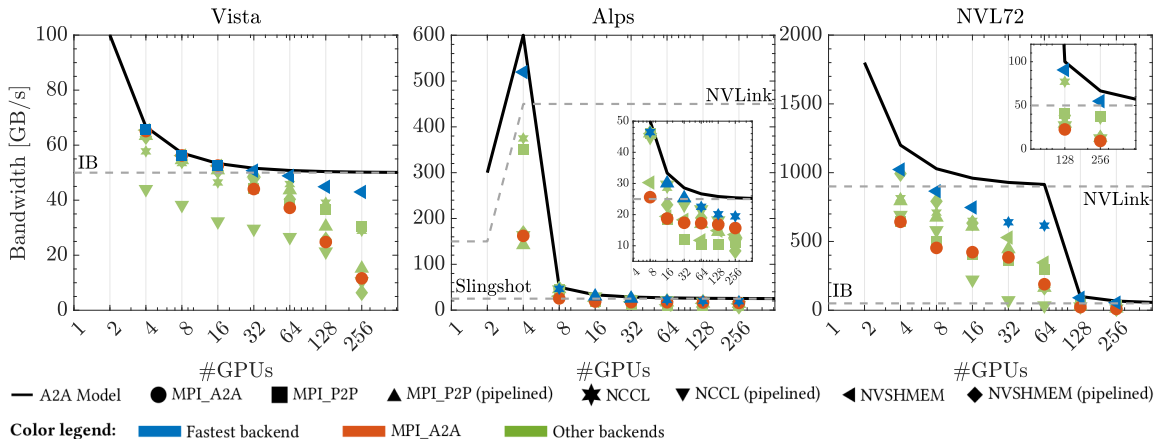
# All-to-all Model Validation: Benchmarking Tests

- Fixed p2p message size for every 8x increase in #GPUs.
- At least one A2A backend achieves near peak model performance.
- Model acts as a reasonable upper bound for achievable A2A bandwidth.



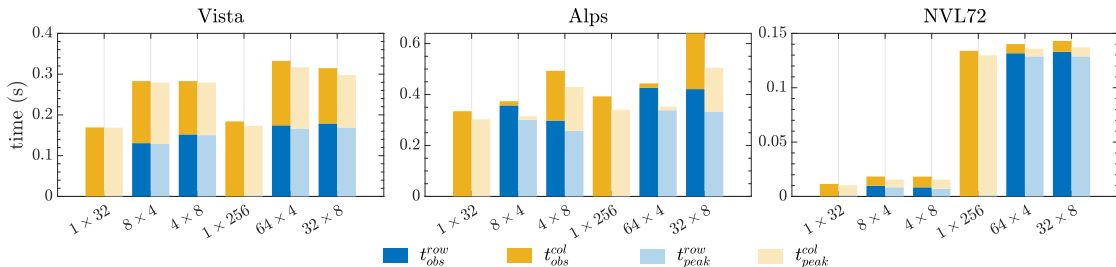
# All-to-all Model Validation: Strong Scaling

- Decreasing p2p message size as #GPUs increases: 8 GiB to 125 MiB.
- Some impact from unmodeled latency terms at the smallest message sizes.



# 2D Distributed Transpose Analysis

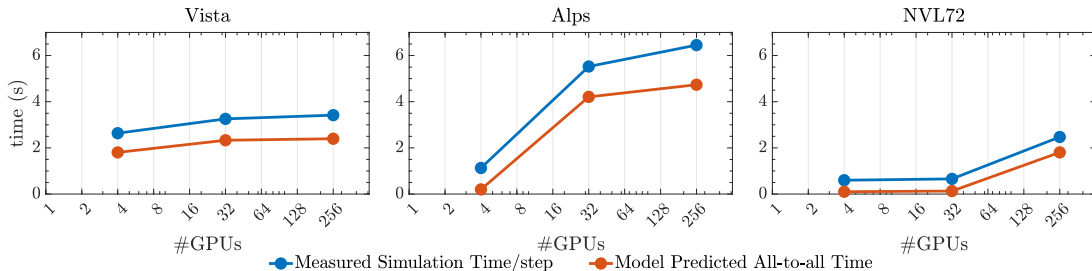
- Measured row (blue) and column (yellow) communicator A2A time (solid color) versus model predictions (faded color) for different  $P_r \times P_c$ .
- Fixed message size  $m$  over  $\#GPUs = 32$  and  $256$ .
- $P_r = 1$  most performant, although significant impact of “near” group size.
- Systems with large “near” groups beneficial for 2D decompositions.



# Weak Scaling of Benchmark Code

What can the model tell us about weak scaling in a communication bandwidth bound application?

- Measured simulation time/step closely mirrors the predicted theoretical best  $t_{a2a}$ .



Weak scaling is not flat: the near and self communication pathways contribute time improvements far beyond the “near” group size  $n$ .



# Summary & Conclusions

## Peak Performance Model

- ✓ Modeling paradigm to obtain an accurate quantitative measure of all-to-all performance given message and system parameters.
- ✓ Accounts for hierarchical structure of modern multi-GPU per node systems with global “far”, tightly connected “near” and local “self” communication pathways.

## Application and Validation

- ✓ Benchmarking tests on three systems validate theoretical upper bound from model as well as model predictions for 2D distributed transpose.
- ✓ “near” and “self” pathways boost communication well beyond “near” group size.
- ✓ No single communication library hits peak all-to-all performance across all process counts and systems.
- ✓ Weak scaling is not flat despite near-peak all-to-all communication performance.