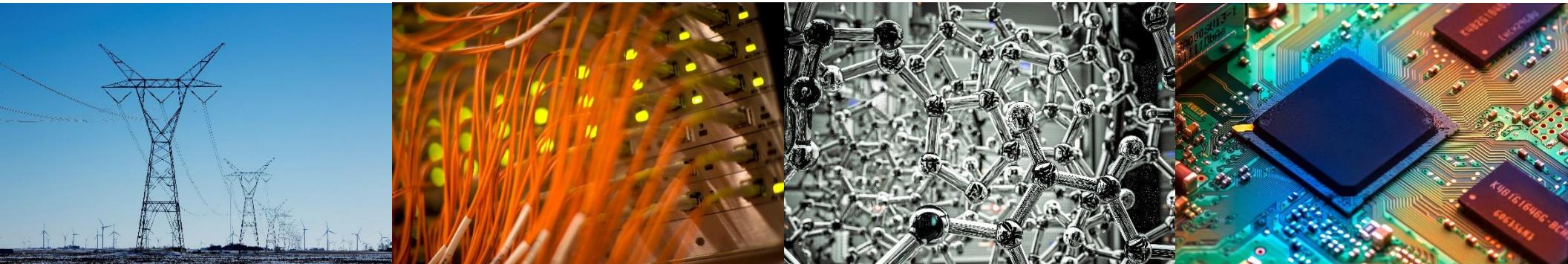


# Comparative Performance Evaluation of Multi-GPU MLFMM Implementation for 2-D VIE Problems

**Carl Pearson**, Mert Hidayetoglu, Wei Ren, Weng Cho Chew, Wen-Mei Hwu  
University of Illinois Urbana-Champaign



# Outline

This work: compare MLFMM performance on two systems

Brief introduction to **Multilevel Fast Multipole Method**

## Some Implementation Notes

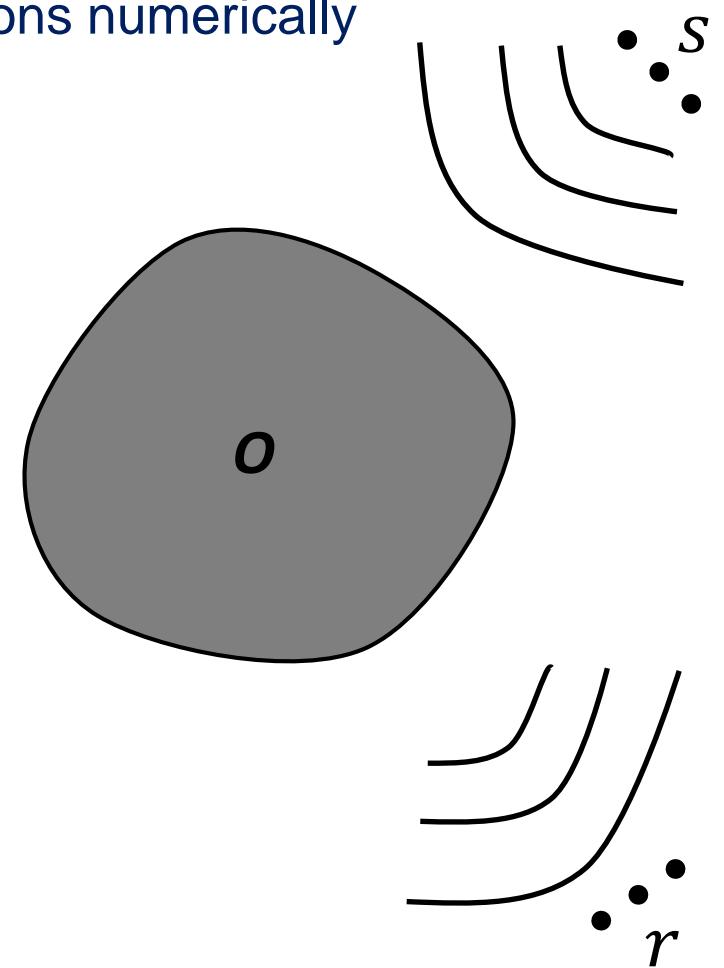
- Matrix Representation of MLFMM Operations
- Kernels and Optimizations
- MPI Parallelization
- Overlapping Communications with Computations

## Results

- Blue Waters and IBM S822LC

# MLFMM

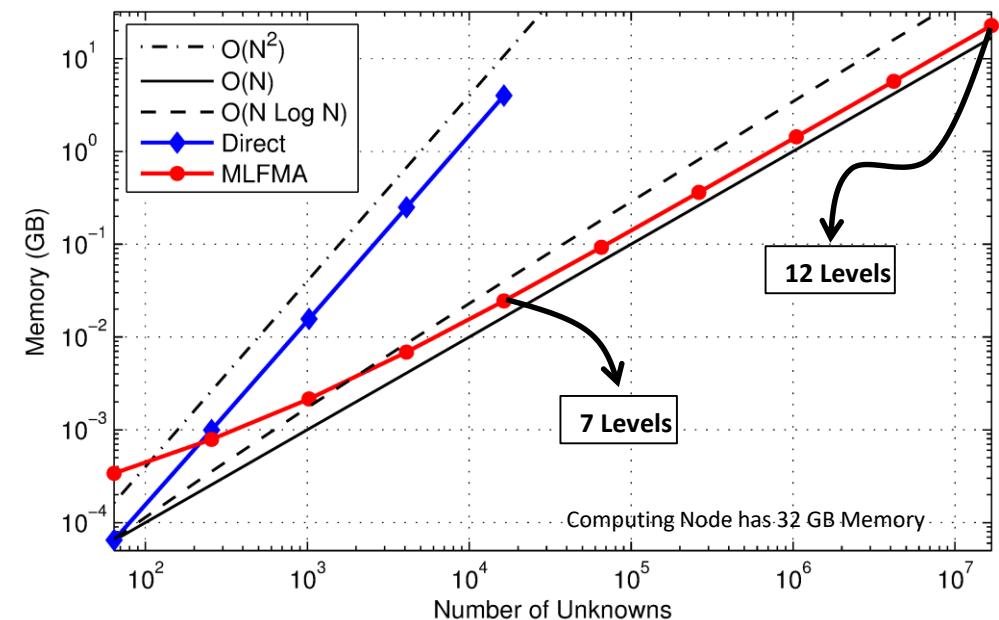
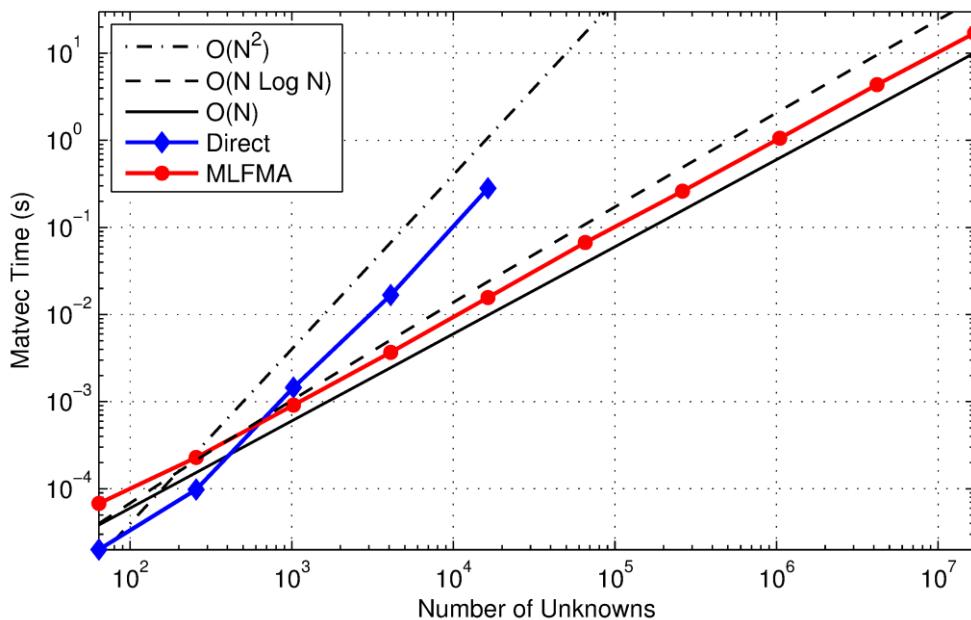
- Solve large scale electromagnetic wave equations numerically
  - Electromagnetics
  - Acoustics
  - Geophysics
  - Radar Cross Section Calculations
  - Medical Imaging
  - Radar Imaging
  - Antenna Design



# Algorithmic Speedup with MLFMM

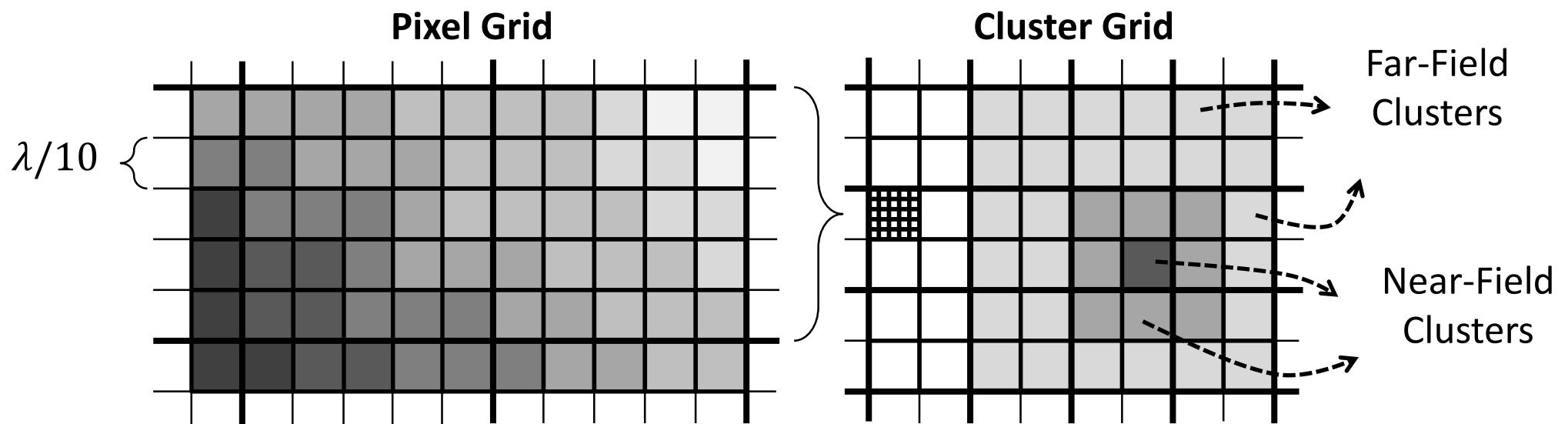
- Direct Methods:  $\mathcal{O}(N^3)$
- Iterative Methods:  $\mathcal{O}(N^2)$
- Fast Multipole Method:  $\mathcal{O}(N^{1.4}) - \mathcal{O}(N^{1.5})$
- Multilevel Fast Multipole Algorithm:  $\mathcal{O}(N) - \mathcal{O}(N \log N)$

$$\begin{aligned} b &= \bar{A}x && \text{known} \\ x &= \bar{A}^{-1}b && \text{unknown} \end{aligned}$$

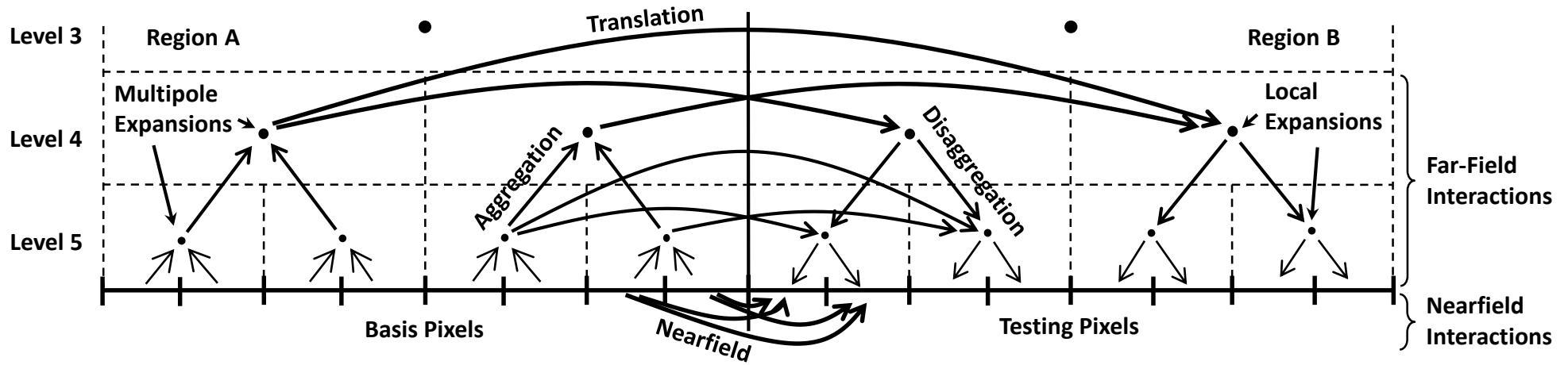


MLFMM has desirable time and memory scaling characteristics

# From Approximate to Full-Wave



# MLFMM Schematic



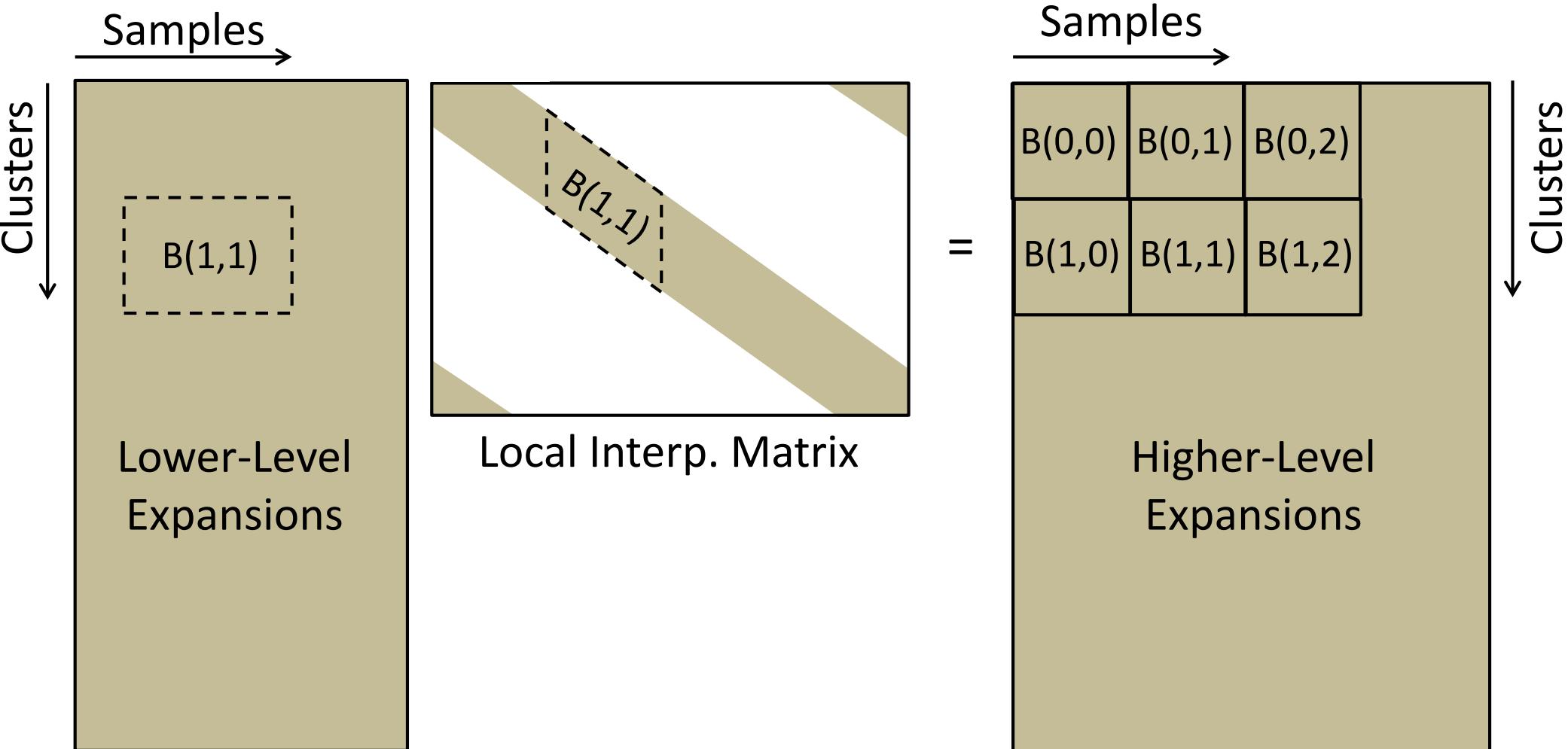
## MLFMA Operation

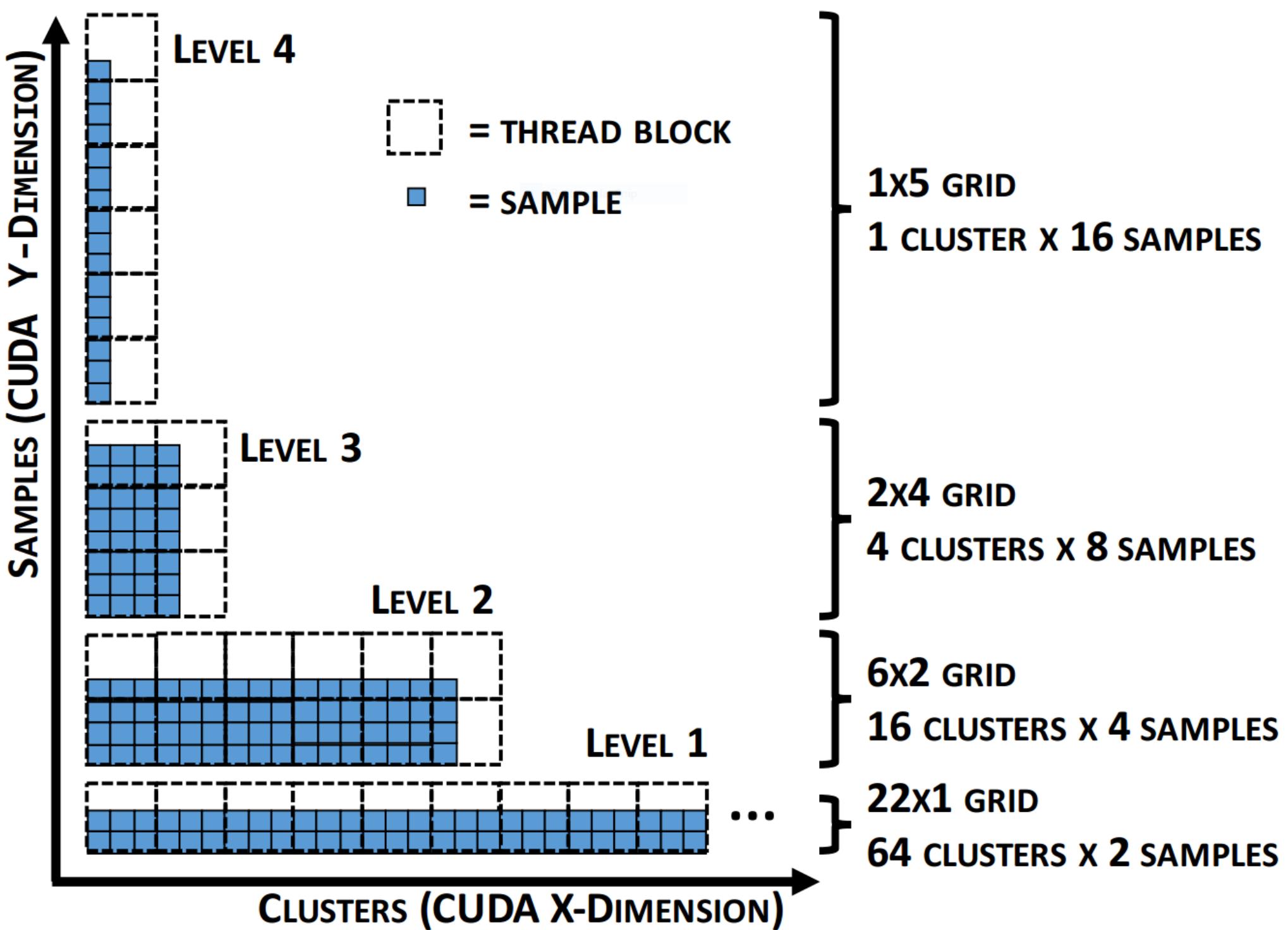
- Multipole & Local Expansions
- Interpolations & Anterpolations
- Multipole & Local Shiftings
- Translations
- Near-Field Interactions

## Structure

- Dense
- Band-Diagonal
- Diagonal
- Diagonal
- Sparse

# Matrix Formulation of Interpolation





# Kernels and Optimizations

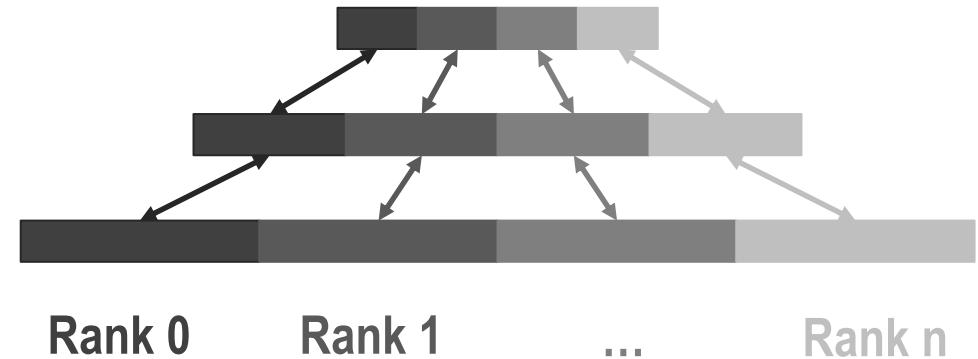
Kernel	Mnemonic	MLFMM Operation
P2M	Particle-to-multipole	Aggregation
M2M	Multipole-to-multipole	
M2L	Multipole-to-local	Translation
L2L	Local-to-local	Disaggregation
L2P	Local-to-particle	
P2P	Particle-to-particle	Nearfield

Traditional GPU optimizations:  
Shared-memory / register tiling  
Thread coarsening

# MLFMM Tree Parallelization

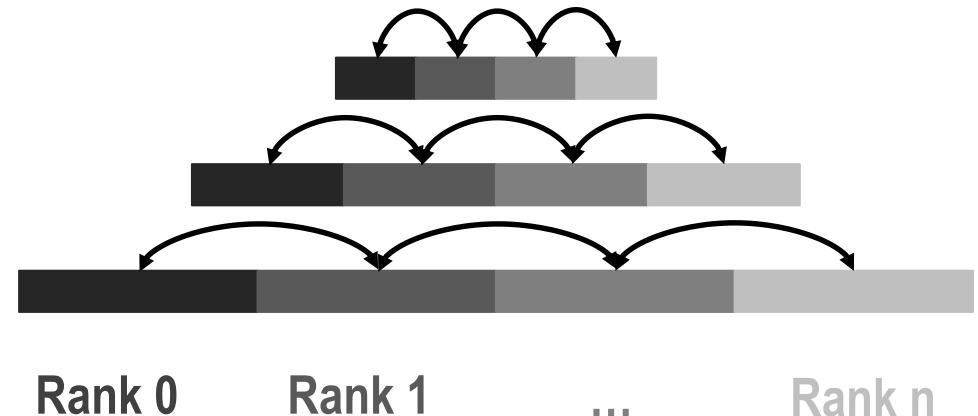
## Aggregation / Disaggregation

- No communication



## Translation

- Inter-rank communication



# Execution Environments

Blue Waters, National Petascale Computing Facility, University of Illinois

22,500 XE CPU and 4,200 XK GPU nodes

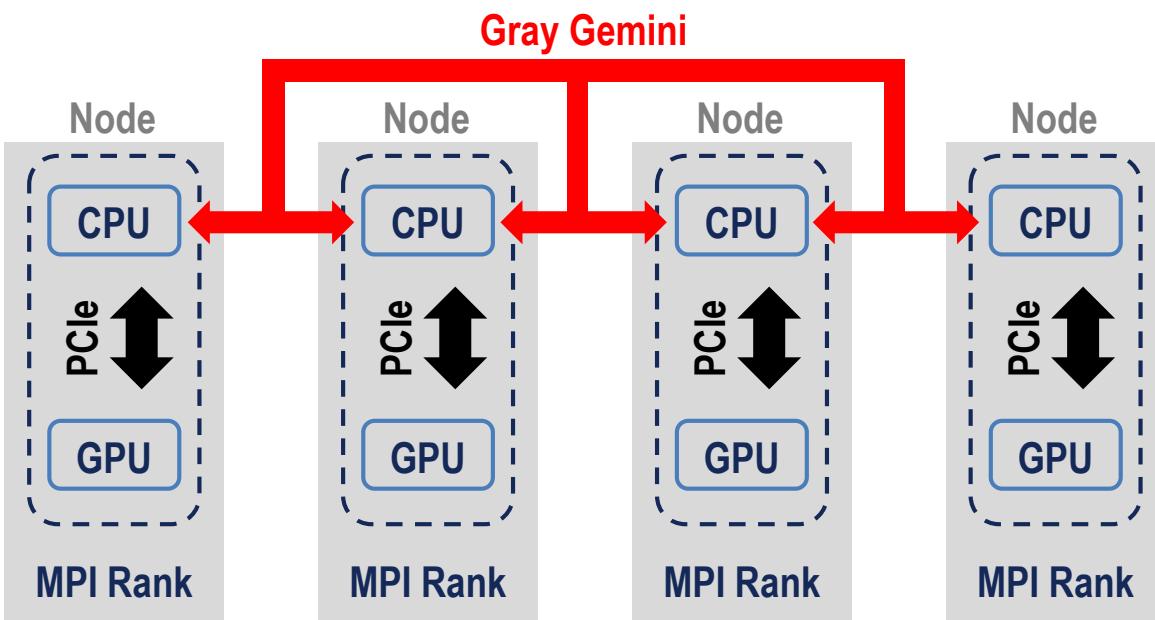
1.5 PB RAM, 13.34 PF peak performance

IBM S822LC “Minsky” system

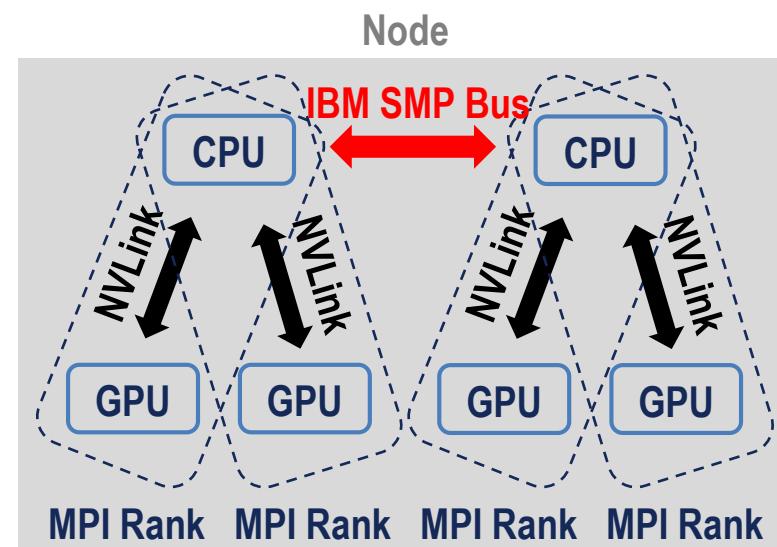
	Blue Waters XK	Blue Waters XE	S822LC “Minsky”
CPU 0	Opteron 6276	Opteron 6276	Power8
CPU 1	--	Opteron 6276	Power8
GPU 0	NVIDIA K20X (Kepler, 6GB)	--	NVIDIA P100 (Pascal, 16 GB)
GPU 1	--	--	NVIDIA P100 (Pascal, 16 GB)
GPU 2	--	--	NVIDIA P100 (Pascal, 16 GB)
GPU 3	--	--	NVIDIA P100 (Pascal, 16 GB)
RAM	32 GB	64 GB	512 GB

# MPI Rank Arrangement

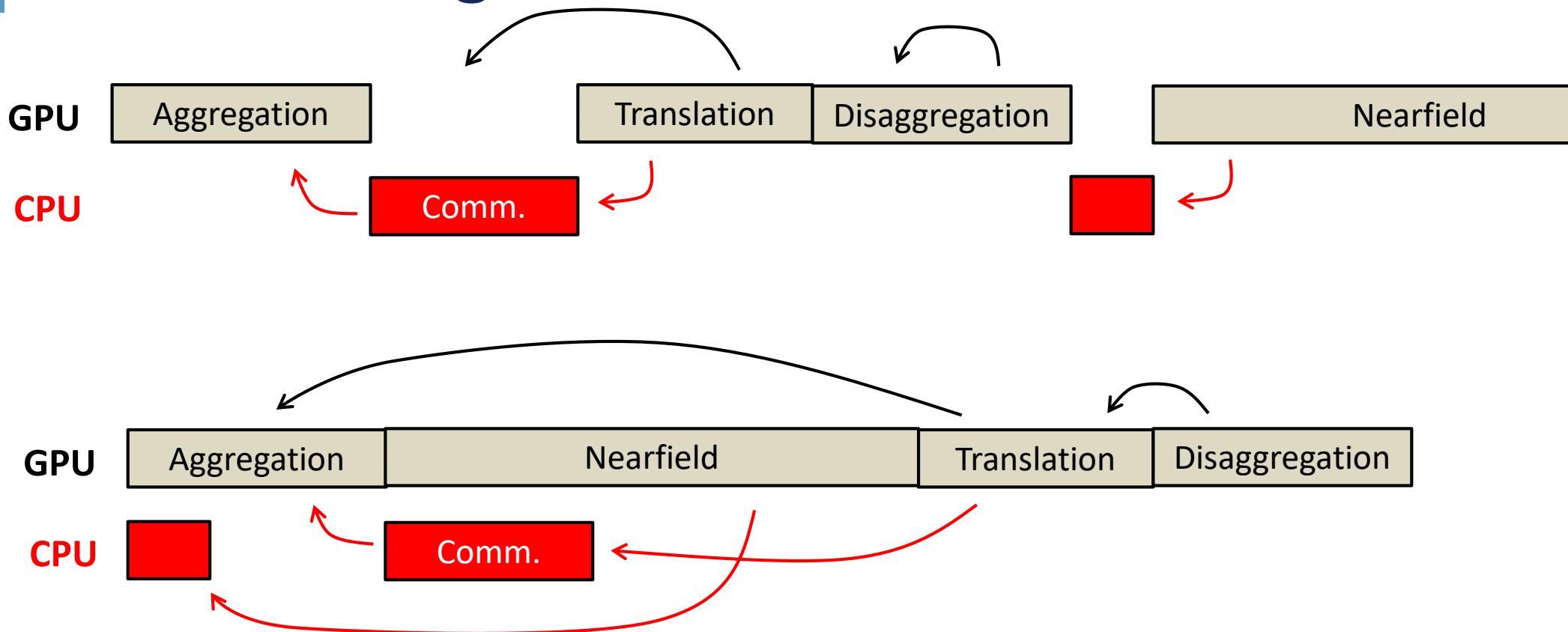
Blue Waters



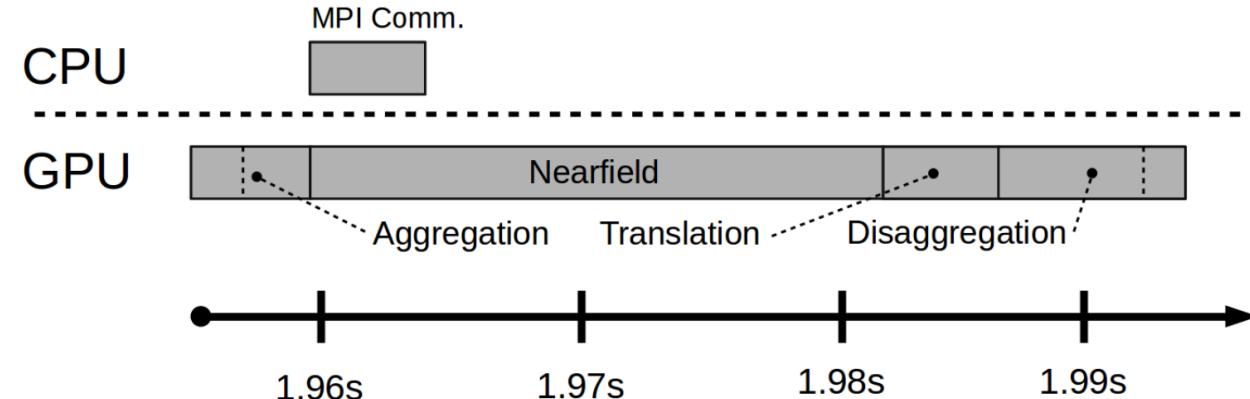
S822LC “Minsky”



# Eliminating Communication Overhead



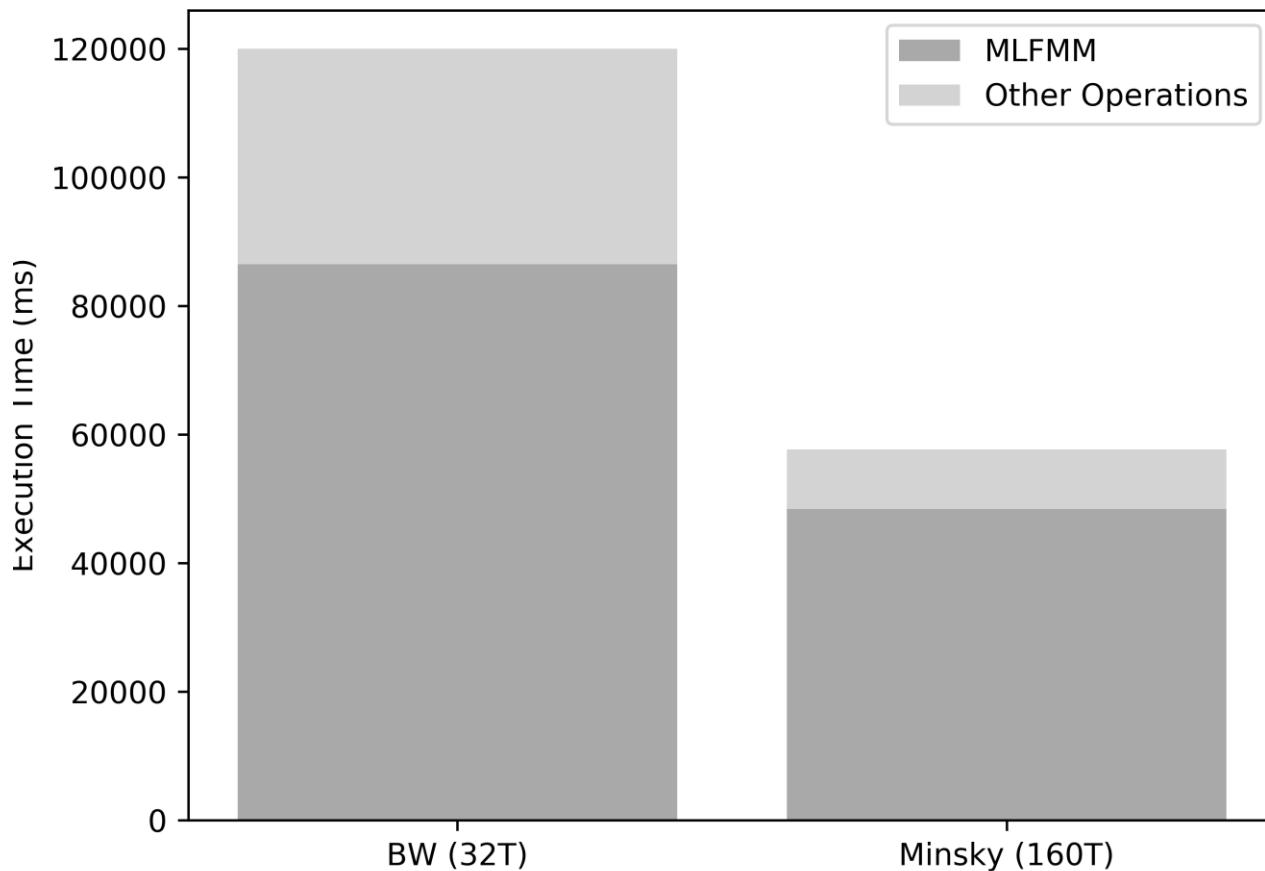
# Multi-Rank MLFMM GPU Scaling



Execution Setup	MLFMM Time (ms)
Blue Waters (1 MPI Rank)	619
Blue Waters (4 MPI Rank)	156 (3.96x)
Blue Waters (16 MPI Rank)	40 (15.58x)
Minsky (1 MPI Rank)	118.92
Minsky (4 MPI Rank)	30.54 (3.89x)

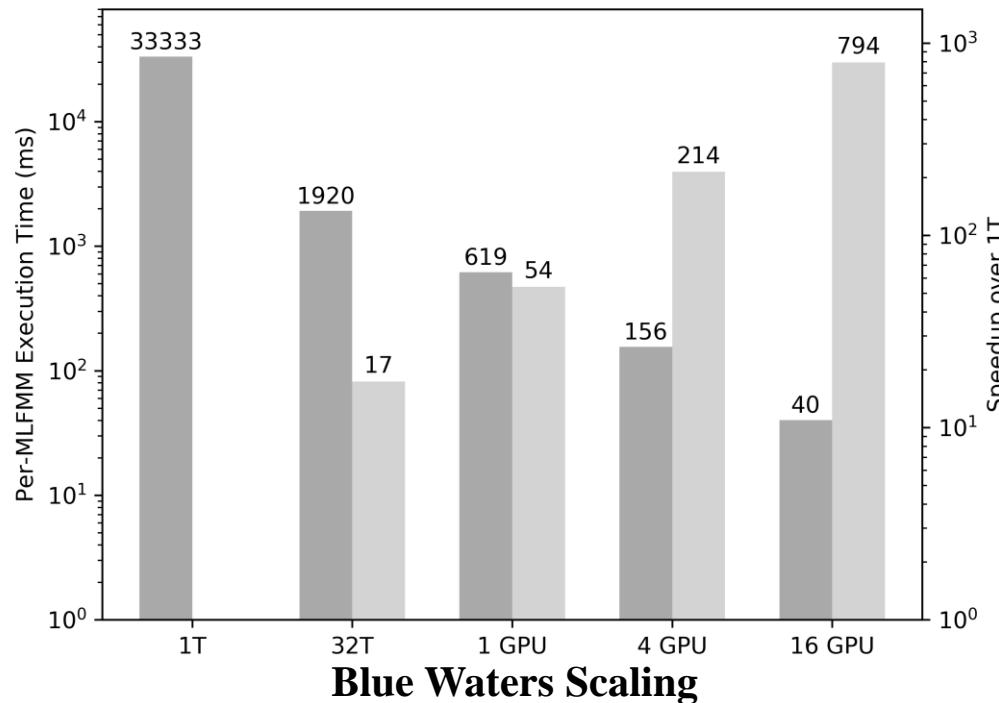
Scalability limits not reached at 16 nodes for us...  
...but in the future, fat nodes can reduce communication costs and improve scalability

# Application Time in MLFMM

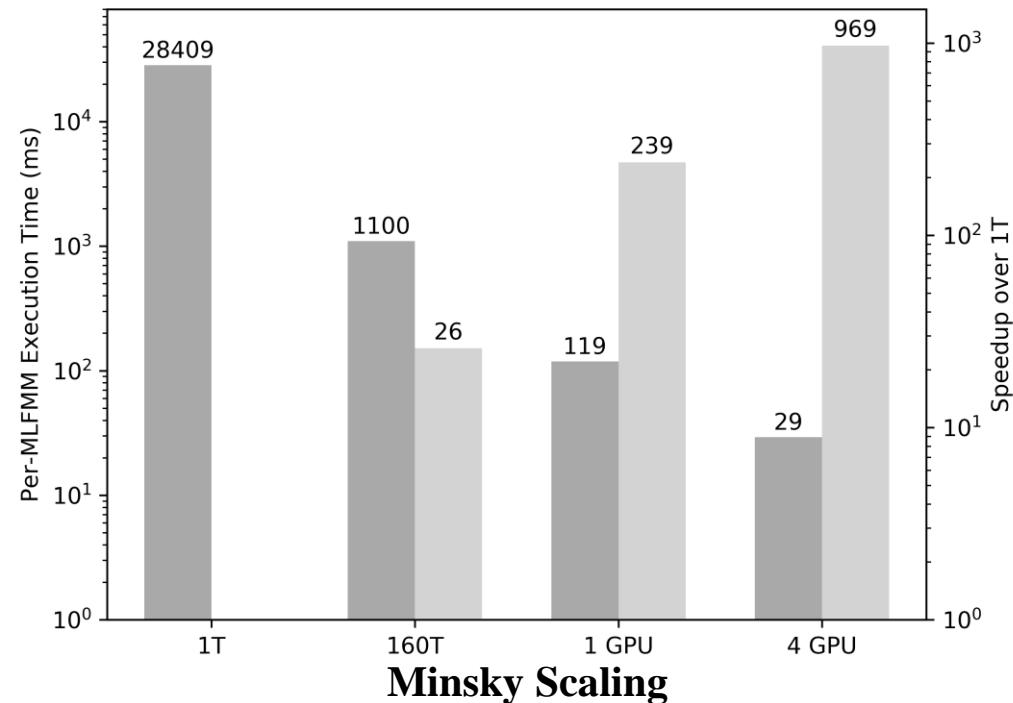


MLFMM is the majority of the execution time

# MLFMM CPU/GPU Scaling



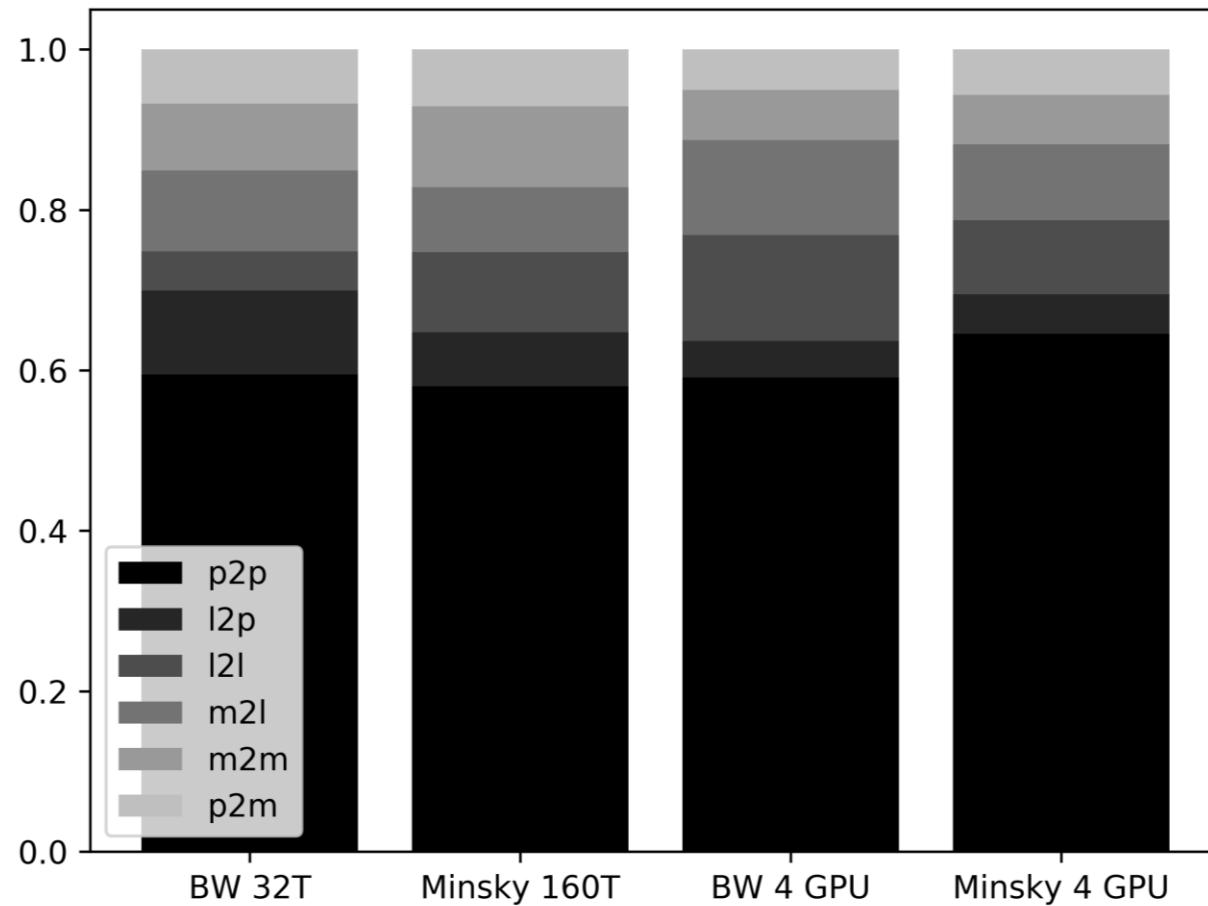
1T → 32T: 17x (16 fp units)  
32T → 1 GPU: 3.1x



1T → 160T: 25x (20 fp units)  
160T → 1 GPU: 9.5x

Minsky CPU Speedup over BW: 1.8x  
Minsky 4-GPU Speedup over BW: 5.3x  
Minsky Node Speedup over BW: 21x

# MLFMM Kernel Breakdown



P2P is most of the execution time  
L2L has largest speedup from K20x to P100

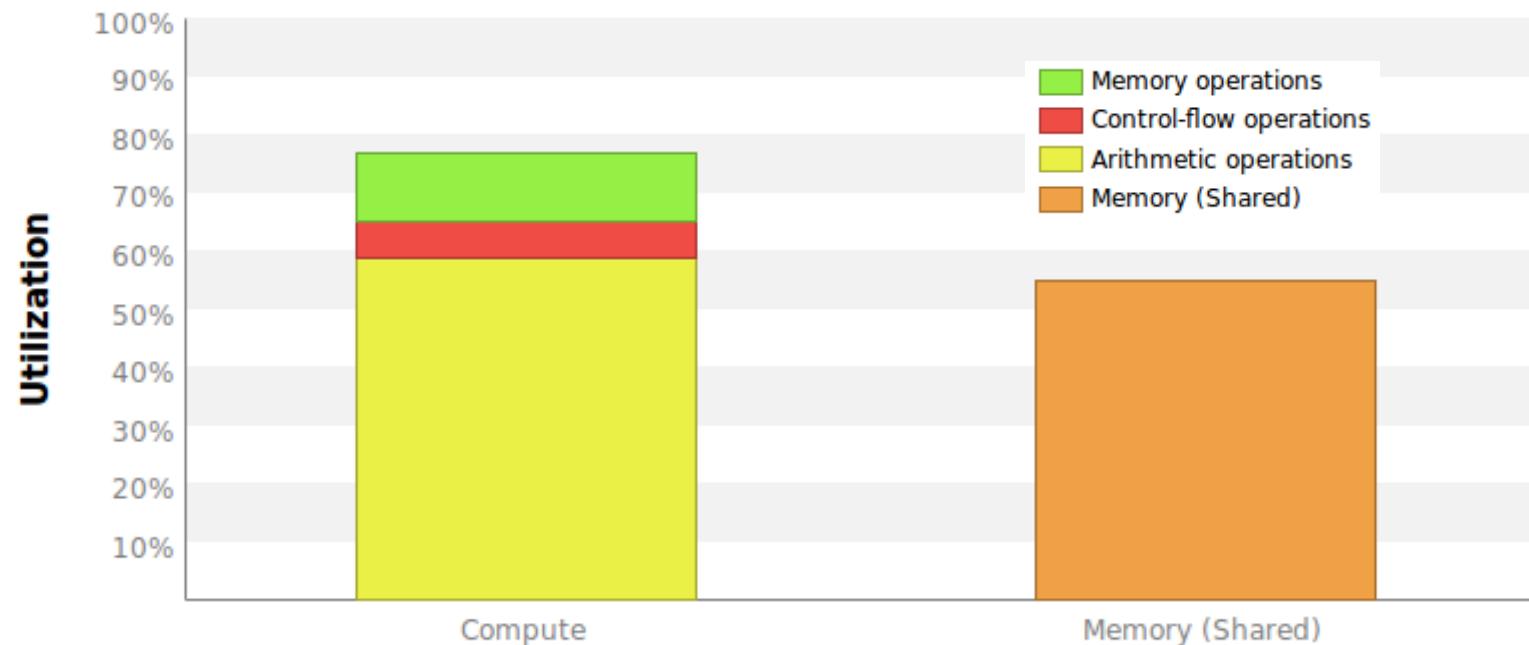
# Kepler vs. Pascal

Feature	K20x Kepler(GK110)	P100 Pascal ( GP100)
Core Clock	732 MHz	1328 MHz
Global Memory Bandwidth	250 GB/s	720 GB/s
Peak GFLOPs(single / double)	3935 /1312	9519 /4760
L2	1.5 MB	4 MB
# of SMs	14	56
Register File	256 KB	256 KB
L1	48 / 32 / 16 KB	0 KB
Shared Memory	16 / 32 / 48 KB	64 KB
"CUDAcores"	192	64
Max Resident Blocks	16	32

**Pascal: More registers and shared-memory per thread, more warps per SM.**

# Lessons from Nearfield Kernel

Longest-running kernel (60% of MLFMM time)



87% threads inactive in inner loop

mod/div on  $2^x$ : immediate 1.3x speedup

**Important to give the compiler information and understand profiling results**

# Lessons from Disaggregation Kernel

	BW	Minsky	Speedup
L2L Time(ms)	78.5	9.9	8.0
MLFMM Time (ms)	633	118.9	5.3

L2L Kernel	BW	Minsky
Theoretical Occupancy	43.8	56.2
Achieved Occupancy	30.7	42.1

**Occupancy limited by shared memory in both cases**

**Relative performance improves due to increased shared memory size**

# Conclusion

- Low-effort port of GPU MLFMM to fat nodes yields good speedup
- Fat nodes will improve scalability for massively parallel MFLMM
- GPU architectures seem to be moving in a beneficial direction

Thank you  
*pearson@illinois.edu*

# MLFMM GPU Performance Data

Kernel	BW 32T (ms)	BW K20x (ms)	BW Speedup (GPU /32T)	Minsky 160T (ms)	Minsky P100 (ms)	Minsky Speedup (GPU / 160T)	Speedup (P100 /K20x)
P2M	127.1	30.9	4.1	72.1	6.4	11.3	4.8
M2M	156.3	37.3	4.2	102.6	6.6	15.6	5.7
M2L	189.6	72.3	2.6	82.7	10.2	8.1	7.1
L2L	91.6	78.5	1.2	101.6	9.9	10.3	8.0
L2P	196.2	28.0	7.0	68.4	5.5	12.4	5.0
P2P	1117.4	361.9	3.1	590.5	72.5	8.1	5.0
Iteration Time	1962.1	633.0	3.1	1074.8	118.9	9.0	5.3

# MLFMM CPU Performance Data

Step	Blue Waters XE 32T(ms)	Minsky 160T (ms)	Speedup BW 32T / 1T	Speedup Minsky 160T / 1T	Speedup Minsky / BW
P2M	127.1	72.1	17.9	25.2	1.8
M2M	156.3	102.6	20.2	24.5	1.5
M2L	189.6	82.7	12.5	19.2	2.3
L2L	91.6	101.6	34.6	28.4	0.9
L2P	196.2	68.4	11.4	26.6	2.9
P2P	1117.4	590.5	18.5	29.0	1.9
MLFMM	1962.1	1074.8	17.3	25.8	1.8





