# Achieving Computation-Communication Overlap with Overdecomposition on GPU Systems

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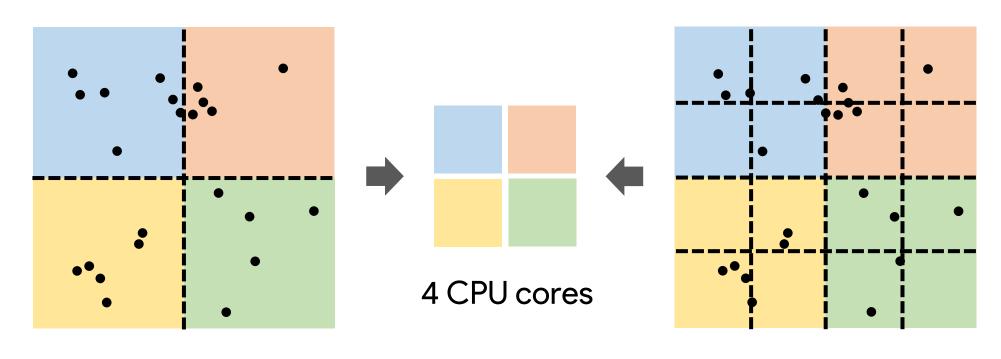
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## Overview

- Increasing gap between single-node computational power and inter-node communication performance on leadership-class systems
   e.g. OLCF Titan → Summit: 1.4 → 40 TFLOPS, 6.4 → 23 GB/s
- Can be tackled from at least 2 directions
  - 1. Improve communication performance itself with software optimizations and better utilization of hardware support (e.g. GPUDirect, SHARP, hardware tag-matching)
  - Reduce impact of communication on overall performance (e.g. communication-avoiding algorithms, computation-communication overlap)
- Focus on achieving computation-communication overlap on GPU systems with overdecomposition using Charm++ (<a href="https://github.com/UIUC-PPL/charm">https://github.com/UIUC-PPL/charm</a>)

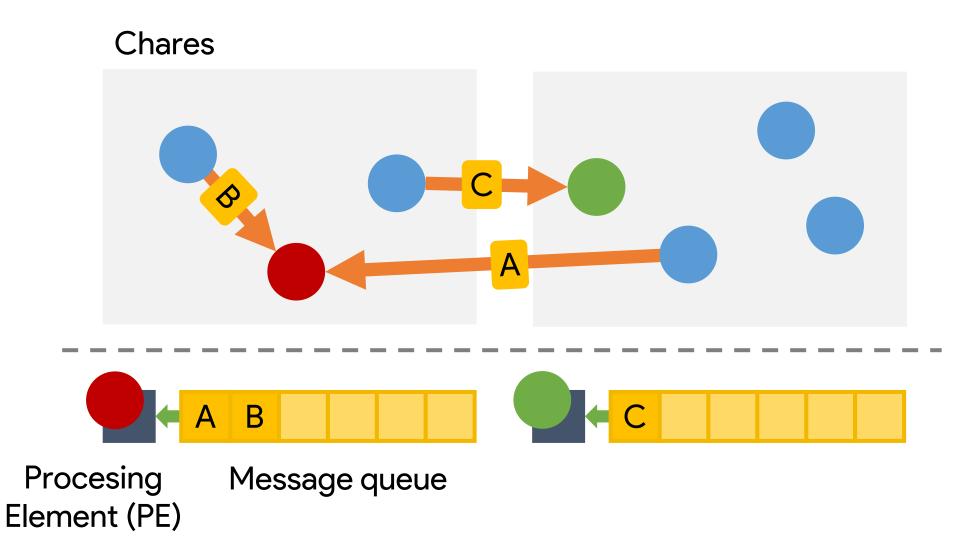
# Overdecomposition



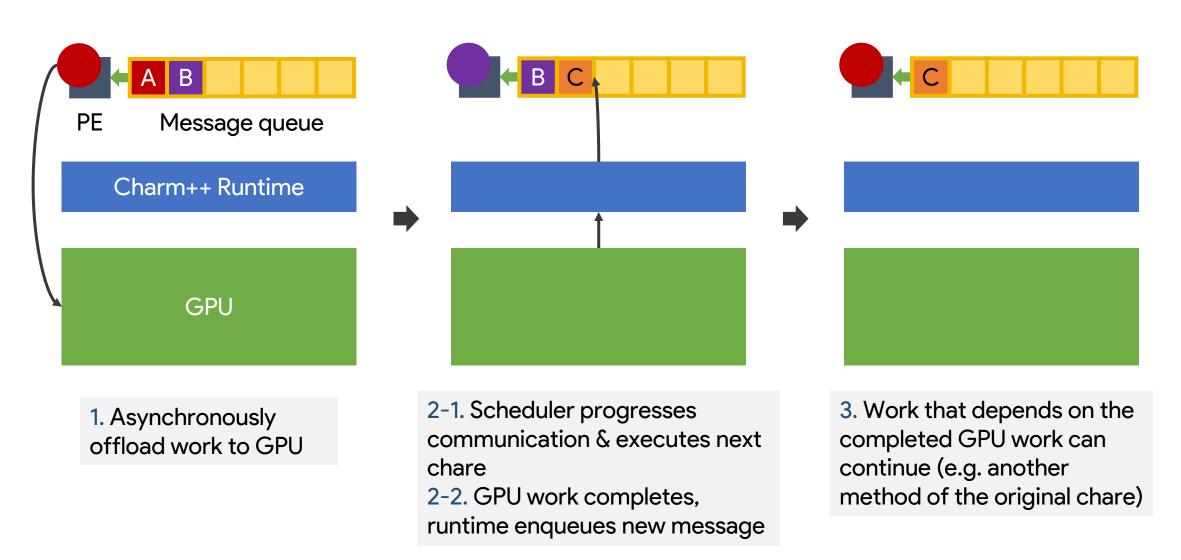
Per-process decomposition (MPI)

Overdecomposition (Charm++)

# Asynchronous Message-Driven Execution



#### **GPU Execution in Charm++**



## Achieving Computation-Communication Overlap

#### 1. Support asynchronous progress in the runtime

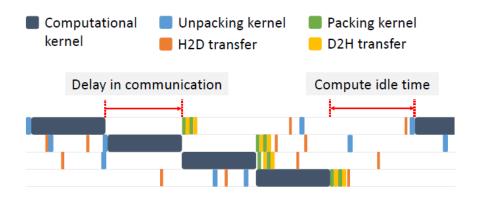
- Avoid synchronization CUDA APIs (e.g. cudaStreamSynchronize)
  - Charm++ scheduler blocked from performing other chares' work
  - Cannot make forward progress on communication (without comm. threads)
- Directly using CUDA async APIs to determine completion is infeasible
  - Scheduler-driven execution in Charm++
  - CUDA-generated thread disassociated from the Charm++ runtime
- hapiAddCallback(cudaStream\_t stream, CkCallback\* callback)
  - Allows user to schedule a Charm++ callback to be invoked when GPU operations complete in the specified CUDA stream
  - Two compile-time configurable mechanisms based on CUDA Callback and CUDA Events (default)
- https://charm.readthedocs.io/en/latest/charm++/manual.html#gpu-support

# Achieving Computation-Communication Overlap

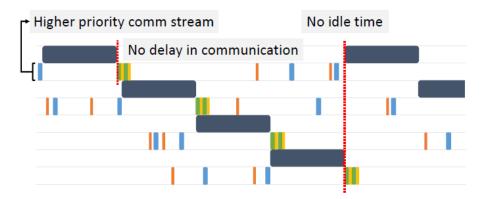
#### 2. Prioritize communication-related GPU operations in the application

- Single CUDA stream per chare: delays in communication-related operations
   (host-device data transfers, packing/unpacking kernels) due to computational kernels offloaded from other chares to the same GPU
- Need separate streams for compute and communication (with higher priority for communication)
- More complex design may be necessary, as for MiniMD (described in paper)

# Achieving Computation-Communication Overlap



(a) Single CUDA stream per chare. Communication is delayed by a computational kernel enqueued from another chare, causing idle time between iterations.



(b) Separate compute/communication CUDA streams per chare, with the communication stream given higher priority. Iterations continue without idle times in between.

Fig. 3. Execution timelines of Jacobi2D with four chares mapped to a single GPU.

#### **Evaluation Platforms**

- OLCF Summit
  - 6 NVIDIA Tesla V100s per node
- LLNL Lassen
  - 4 NVIDIA Tesla V100s per node
- 1 process with 1 PE/core per GPU
  - e.g. 6 PEs and 6 GPUs per compute node on Summit

#### Benchmarks

Iterative proxy apps

#### Jacobi3D

- Jacobi iteration performed on 3D grid, overdecomposed into chares
- Near-neighbor exchange of halo data (up to 6 neighbors)

#### MiniMD

- Proxy app for LAMMPS molecular dynamics code
- Converted MPI + Kokkos to Charm++ + Kokkos
- CUDA-aware MPI converted to explicit host-device transfers and host messages
- Kokkos responsible for computational kernels and host-device data movement
- Neighbor exchange of atoms, Lennard-Jones force calculation

## Performance Results – Jacobi3D

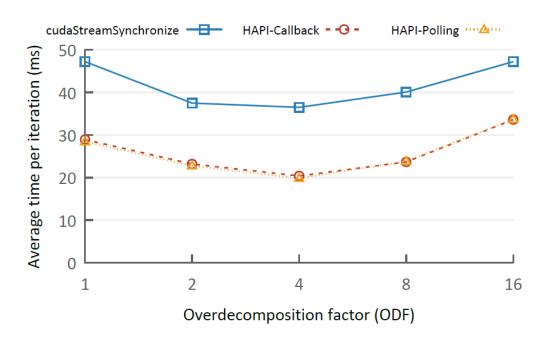
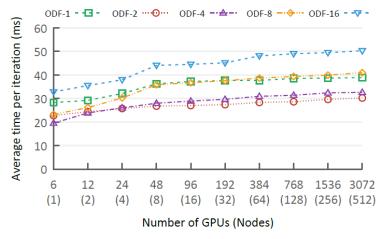
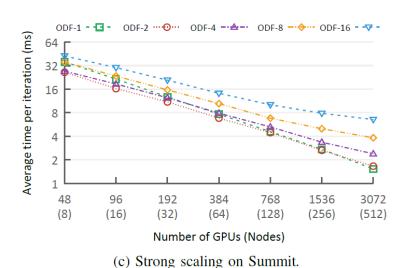


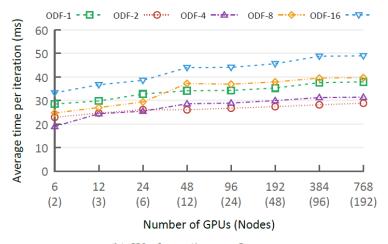
Fig. 5. Performance of Jacobi3D with varying overdecomposition factors on a single node of OLCF Summit.

## Performance Results - Jacobi3D

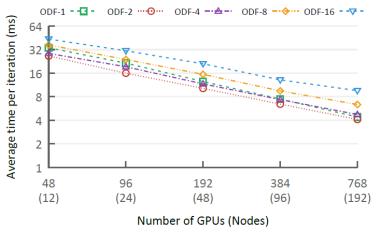


(a) Weak scaling on Summit.





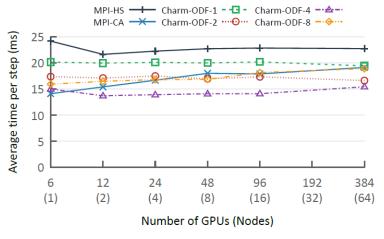
(b) Weak scaling on Lassen.



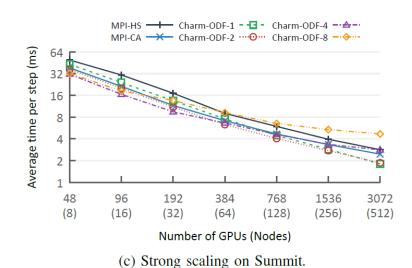
(d) Strong scaling on Lassen.

Fig. 6. Weak & strong scaling performance of Jacobi3D.

## Performance Results - MiniMD

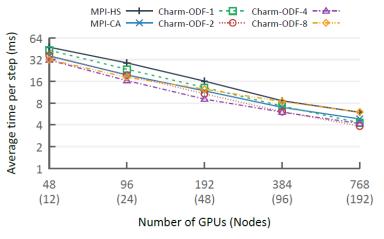


(a) Weak scaling on Summit.



MPI-HS — Charm-ODF-1 - □ Charm-ODF-4 -- △ --MPI-CA - Charm-ODF-2 .... Charm-ODF-8 - ... Average time per step (ms) 10 12 48 192 384 24 96 6 (2)(3) (48)(96)(12)(24)(6)Number of GPUs (Nodes)

(b) Weak scaling on Lassen.



(d) Strong scaling on Lassen.

Fig. 7. Weak & strong scaling performance of MiniMD.

#### Conclusion

- Up to 50% and 47% improvement in overall performance with Jacobi3D and MiniMD, respectively
- With careful design of the application to prioritize communication and support for asynchronous progress of GPU work in the runtime system, overdecomposition can significantly improve performance, especially in weak scaling
- Combining Charm++ (for overdecomposition and communication) with performance portability models (e.g. Kokkos) can be a good path forward for upcoming vendor-heterogeneous GPU systems

# Ongoing & Future Work

- Support for direct GPU-GPU transfer in Charm++
  - Release 6.11: CUDA backend for within-node messaging
  - Exploring UCX, NCCL backends to support inter-node messaging
- Mitigating fine-grained overheads
  - Improve effect of overdecomposition with strong scaling
  - Kernel aggregation
- Dynamic load balancing with GPU loads

# Thank you! Questions?

Please feel free to reach out to jchoi157@illinois.edu.