

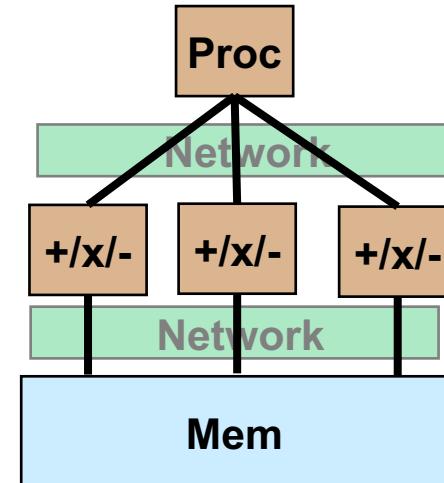
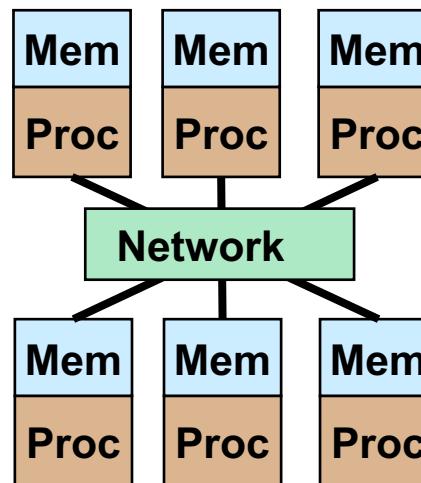
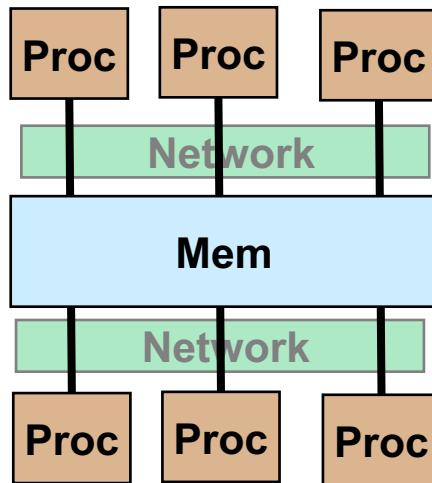
A photograph of a modern architectural building with a long, horizontal glass facade featuring vertical blinds. The building is set against a bright blue sky filled with wispy white clouds. The text is overlaid on the left side of the image.

# Partitioned Global Address Space Languages

## CS267 Lecture 11

Kathy Yelick

# Parallel Machines and Programming



Shared Memory	Distributed Memory	Single Instruction Multiple Data (SIMD)
OpenMP, Threads	MPI (send/receive)	Data parallel (collectives)
Processors execute own instruction stream	Processors execute own instruction stream	One instruction stream (all run same instruction)
Communicate by reading/writing memory	Communicate by sending messages	Communicate through memory
Ideal cost Cost of a read/write is constant	Memory access time depends on size and whether local vs. remote	Assume unbounded # of arithmetic units

# Advantages and disadvantages of each

- Shared memory / OpenMP
  - + **Ease:** Easier to parallelize existing serial code
  - **Correctness:** Race conditions
  - **Scalability:** No locality control for multi-node scaling
  - **Performance:** False sharing, lack of parallelism, etc.
- Message Passing / two-sided MPI
  - **Ease:** More work up front to partition data
  - + **Correctness:** Harder to create races (deadlocks are more of a problem)
  - + **Scalability:** Effectively unlimited
  - + **Performance:** More transparent, but messages are expensive (need to pack/unpack)

# Parallel Programming Problem: Histogram

- Consider the problem of computing a histogram:
  - Large number of “words” streaming in from somewhere
  - You want to count the # of words with a given property

- Shared memory
  - Lock each bucket

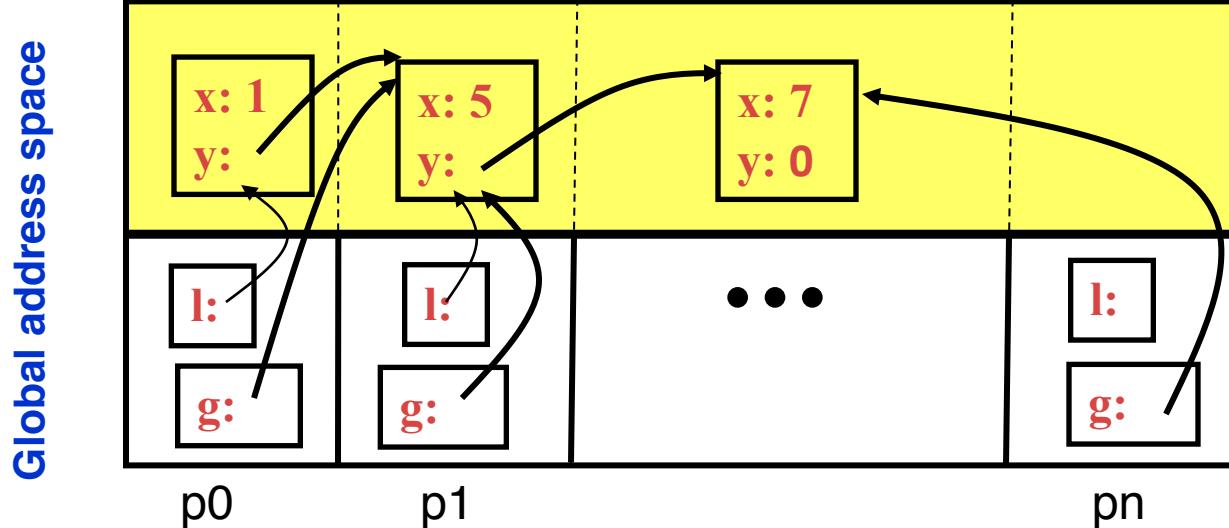


- Message passing: the array is huge and spread out
  - Each processor has a substream and sends +1 to the appropriate processor...
  - When does that processor “receive”?

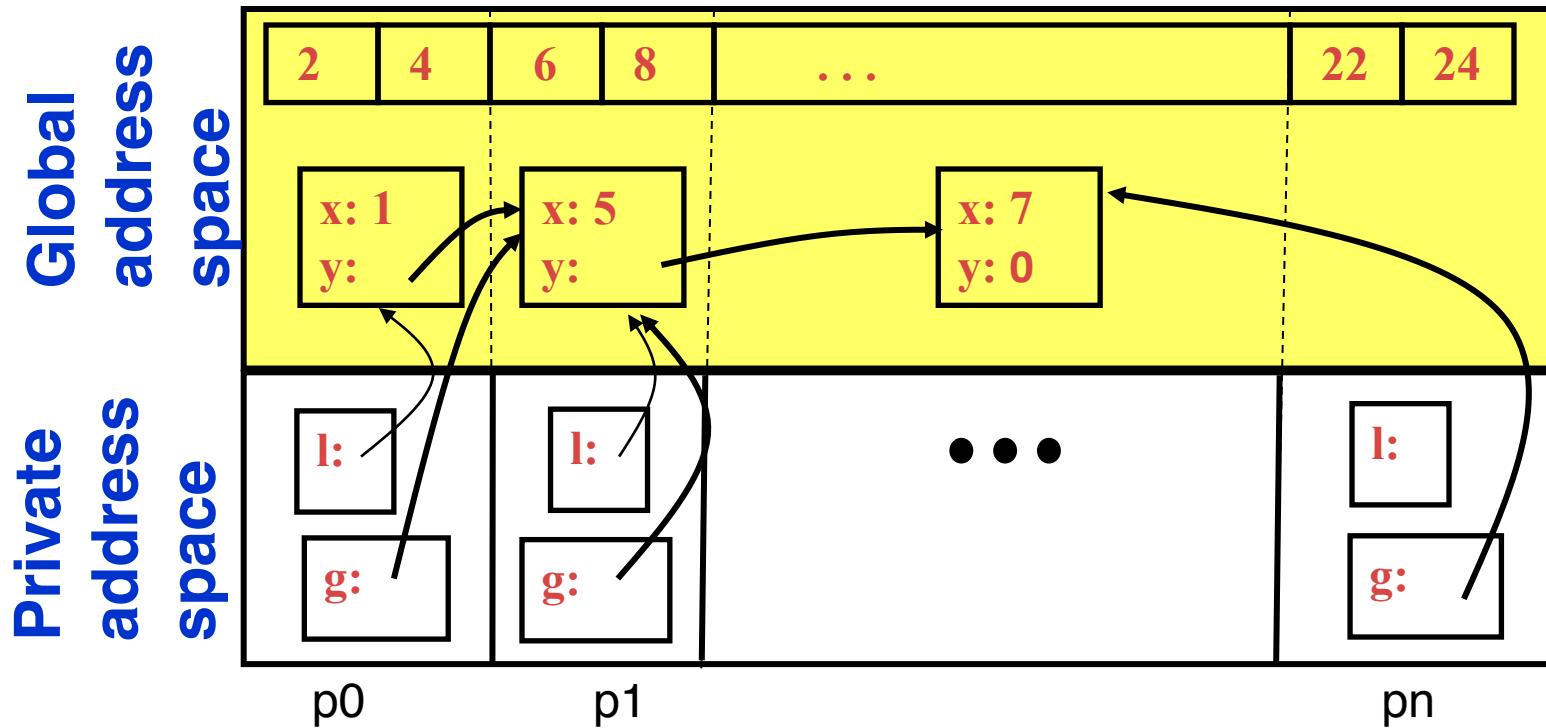


# PGAS = Partitioned Global Address Space

- **Global address space:** thread may directly read/write remote data
  - Convenience of shared memory
- **Partitioned:** data is designated as local or global
  - Locality and scalability of message passing

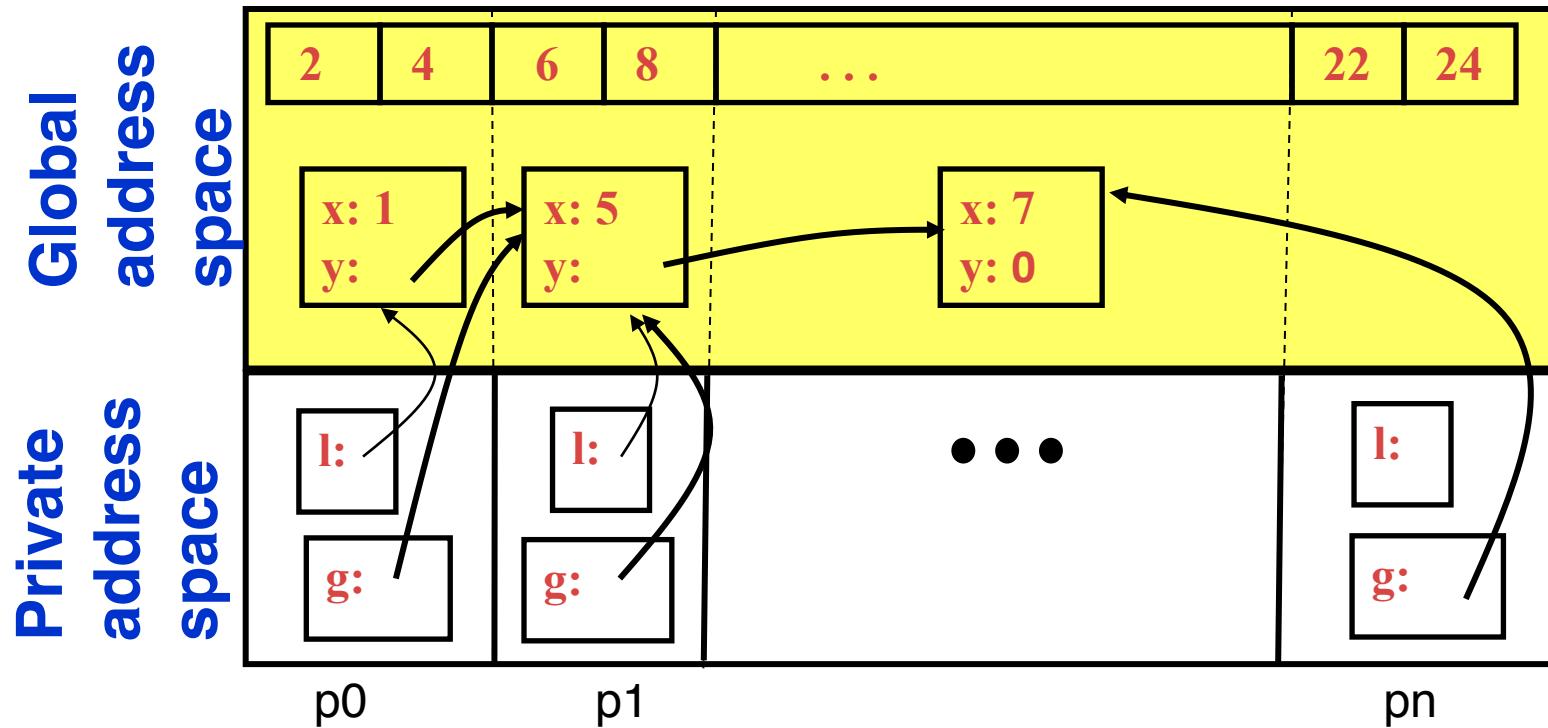


# Partitioned Global Address Space (PGAS)



- Need a way to name remote memory (UPC syntax)
  - Global pointers: `shared int * p = upc_malloc(4);`
  - Distributed arrays: `shared int a [12];`

# One-sided communication in PGAS



- Directly read/write remote memory; partitioned for locality
- One-sided communication underneath (UPC syntax):

Put: `a[i] = ... ; *p = ...; upc_mem_put(..)`

Get: `... = a[i]...; ... = *p; upc_mem_get(...)`

# Goals of PGAS Programming

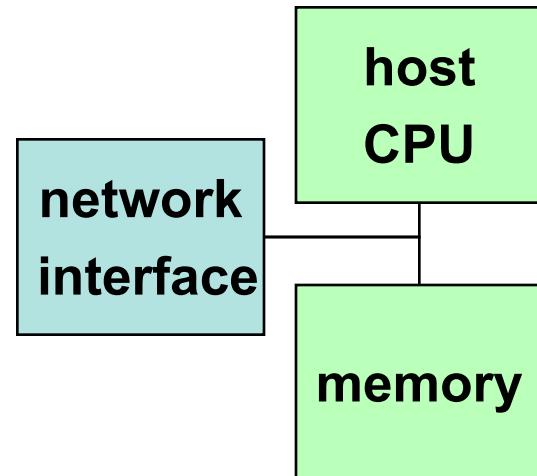
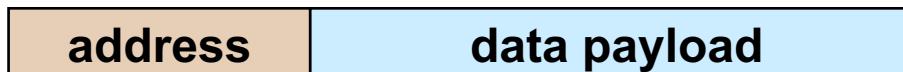
- Applications: convenient programming of irregular codes
  - Graphs
  - Hash tables
  - Sparse matrices
  - Adaptive (hierarchical) meshes
- Machines: expose best available performance on a given machine
  - Low latency for small messages
  - High bandwidth even for medium sized messages
  - High injection rate (number of messages / second)
  - Minimize software overhead and match hardware

# One-Sided Communication in PGAS (e.g., GASnet inside)

**two-sided message**

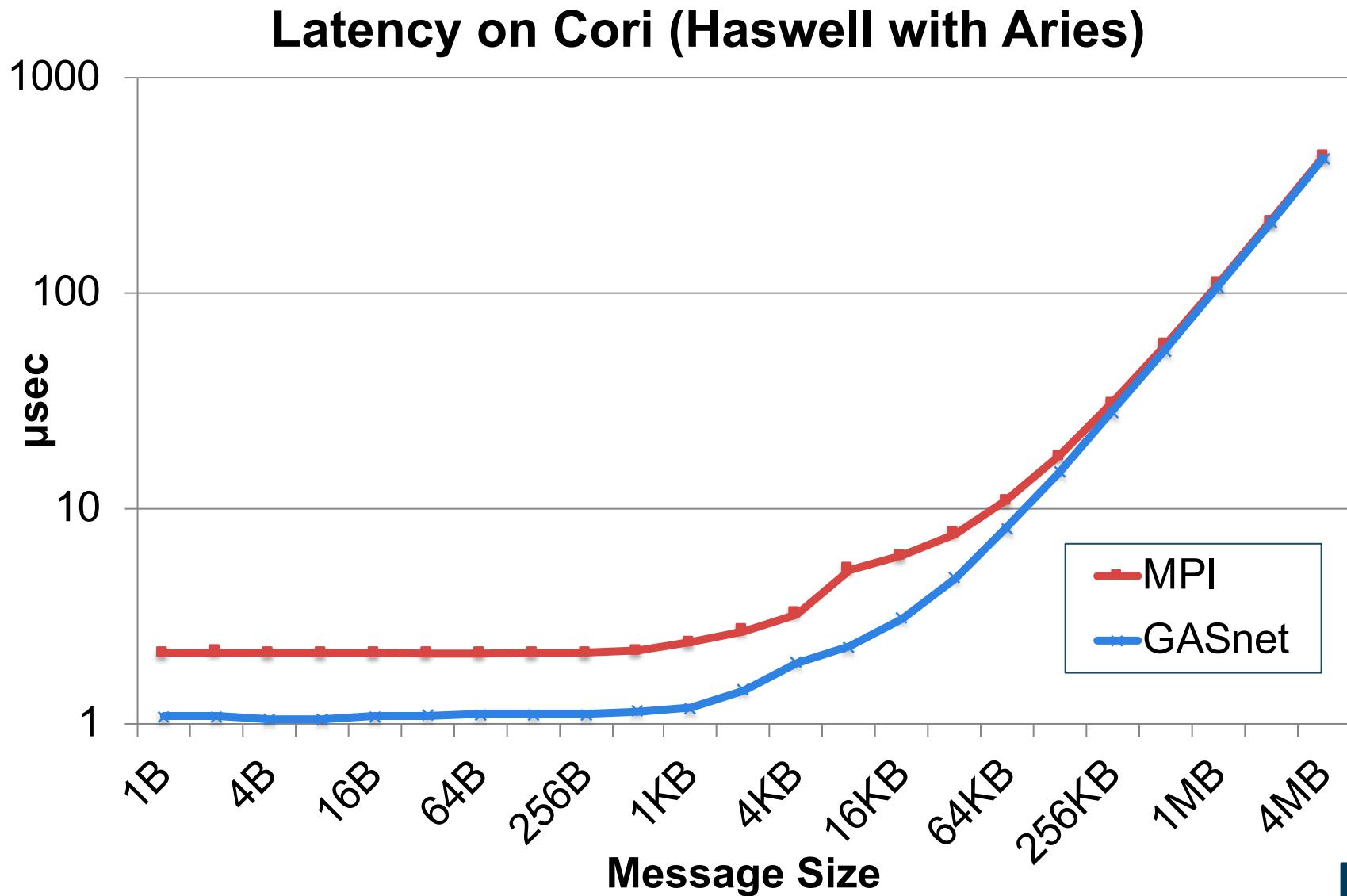


**one-sided put message**

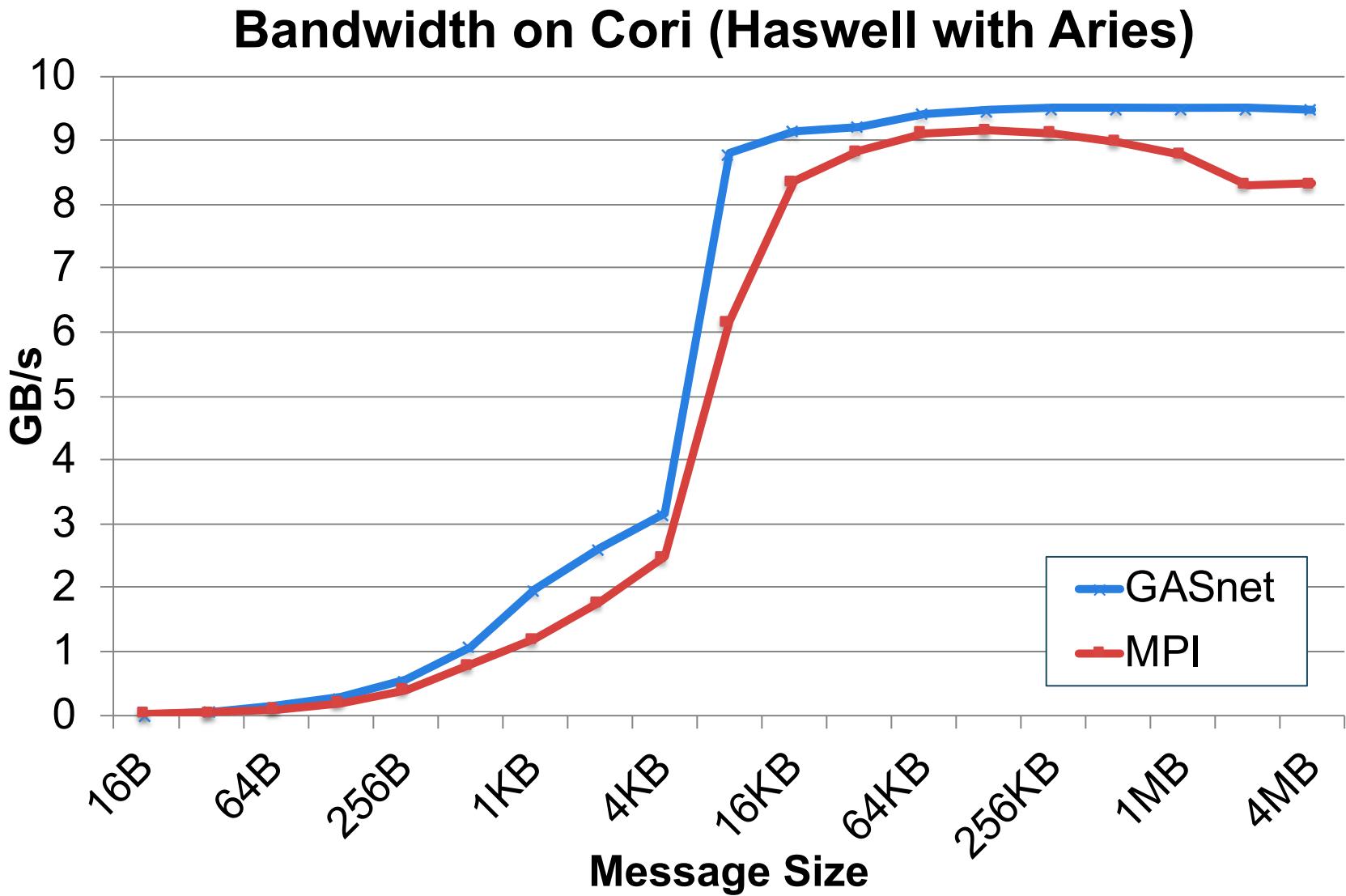


- A two-sided message needs to be matched with a receive
  - Ordering requirements can also hinder bandwidth
- A put/get can be handled by a network interface with RDMA
  - Decouples transfer from synchronization
  - Avoids interrupting the CPU or getting address from CPU
- Will still want to pack data and overlap
  - Pipeline messages, overlap with computation

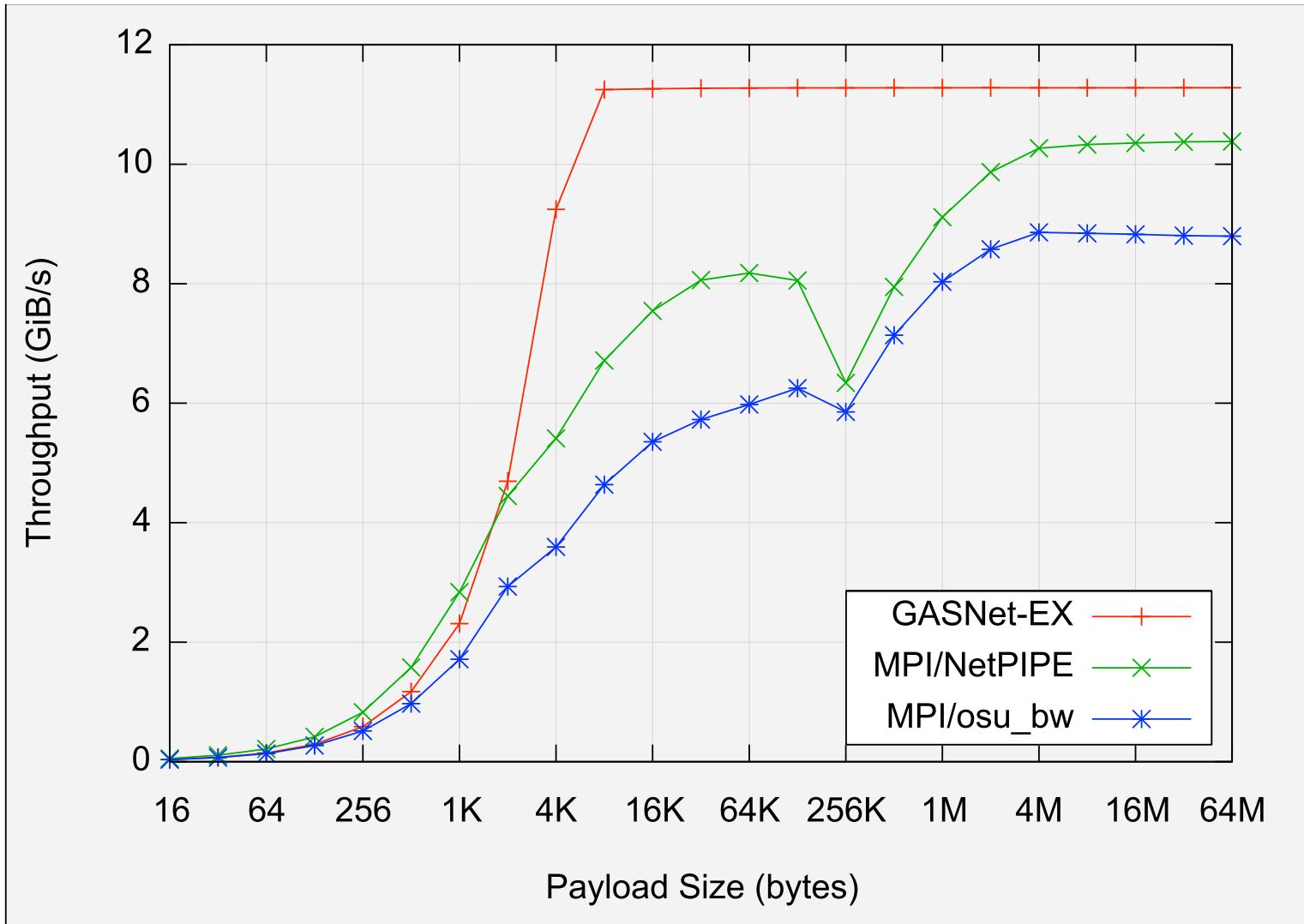
# Latency on a Cray Aries (NERSC Cori-P1)



# Bandwidth on a Cray Aries (NERSC Cori-P1)



# Bandwidth on Infiniband (Summit-Dev at ORNL)



Flood Bandwidth Graphs report tests of achievable one-way bandwidth for point-to-point data transfer between two nodes. All bandwidths have been converted for uniform reporting in units of Gibibytes/sec (GiB/sec), where GiB =  $2^{30}$  bytes.



## ***Introduction to UPC++***

**Led by Scott B. Baden and Paul Hargrove (LBNL)**  
**Slides and examples from Amir Kamil**

**See also the Programmer's Guide at:**

<https://bitbucket.org/upcxx/upcxx/downloads/upcxx-guide-2018.1.0.pdf>

# Hello World in UPC++

- UPC++ programs are just C++ programs
- They run as SPMD programs (similar to MPI)
- They use the GASNet library for communication

```
#include <upcxx/upcxx.hpp>
#include <iostream>
using namespace std;

int main(int argc, char *argv[])
{
    upcxx::init(); // setup UPC++ runtime
    cout << "Hello from " << upcxx::rank_me() << endl;
    upcxx::finalize(); // close down UPC++ runtime
    return 0;
}
```



# Roadmap of the C/C++ PGAS languages

## Early C-based languages

Split-C  
(UCB)

AC (IDA)

PCP (LLNL)

## UPC standard, multiple compilers

Cray UPC

Berkeley  
UPC (LBNL)

GCC UPC  
(Intrepid)

Clang UPC  
(LBNL)

## C++-based languages

Co-Array  
C++ (Cray)

UPC++ v0  
(LBNL)

UPC++ v1  
(LBNL)

Libraries, no  
special syntax

There are also languages Co-Array Fortran (now in standard Fortran), Chapel, Titanium (Java), Legion and more



# UPC++ V1.0 Overview

- A complete redesign of UPC++ that leverages GASNet-EX to deliver better performance and scalability
- A “compiler-free” approach for PGAS
  - Leverage C++ standards and compilers
  - Influence future directions of the C++ standard
- Interoperates with existing programming systems
  - 1-to-1 mapping between MPI rank and UPC++ rank
  - OpenMP and CUDA can be easily mixed with UPC++ in the same way as MPI+X
- Design philosophy:
  - All communication is explicit
  - Most operations are non-blocking
  - No non-scalable data structures

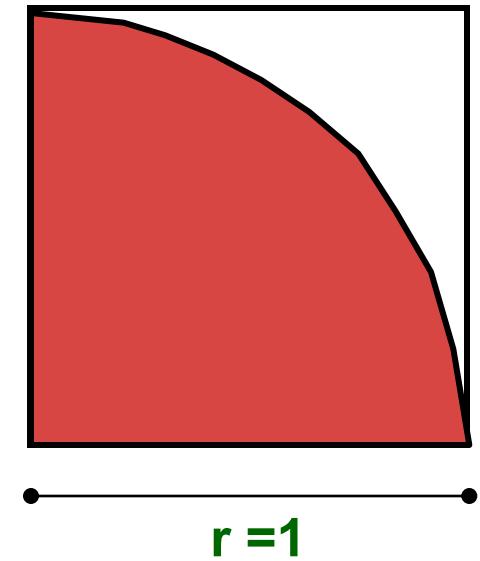


# The API

- Foundational types
  - Global Pointers
  - Futures (and Promises)
  - Distributed Objects
- Communication
  - 1-sided Communication
    - rput/rget (bulk and single element), non-contiguous transfers, memory kinds
  - RPC (remote procedure call)
- Callbacks
- Remote Atomics
- Teams (mechanism for grouping ranks together)
- Progress and the Memory Model

# Example: Monte Carlo Pi Calculation

- Estimate Pi by throwing darts at a unit square
- Calculate percentage that fall in the unit circle
  - Area of square =  $r^2 = 1$
  - Area of circle quadrant =  $\frac{1}{4} * \pi r^2 = \pi/4$
- Randomly throw darts at x,y positions
- If  $x^2 + y^2 < 1$ , then point is inside circle
- Compute ratio:
  - # points inside / # points total
  - $\pi = 4 * \text{ratio}$



# Pi in UPC++

- Independent estimates of pi:

```
int main(int argc, char **argv) {
```

```
    upcxx::init();
```

```
    int hits, trials = 0;
```

```
    double pi;
```

Each rank gets its own copy of these variables

```
    if (argc != 2) trials = 1000000;  
    else trials = atoi(argv[1]);
```

Each rank can use input arguments

```
    generator.seed(upcxx::rank_me()*17);
```

Initialize random in math library

```
    for (int i=0; i < trials; i++) hits += hit();  
    pi = 4.0*hits/trials;  
    cout << "PI estimated to " << pi << endl;
```

```
    upcxx::finalize();
```

Each rank calls “hit” separately

# C++ Helper Code for Pi (in C++11)

- Required includes and variables:

```
#include <iostream>
#include <random>
#include <upcxx/upcxx.hpp>
default_random_engine generator;
uniform_real_distribution<> dist(0.0, 1.0);
```

- Function to throw dart and calculate where it hits:

```
int hit() {
    double x = dist(generator);
    double y = dist(generator);
    if (x*x + y*y <= 1.0) {
        return 1;
    } else {
        return 0;
}
```

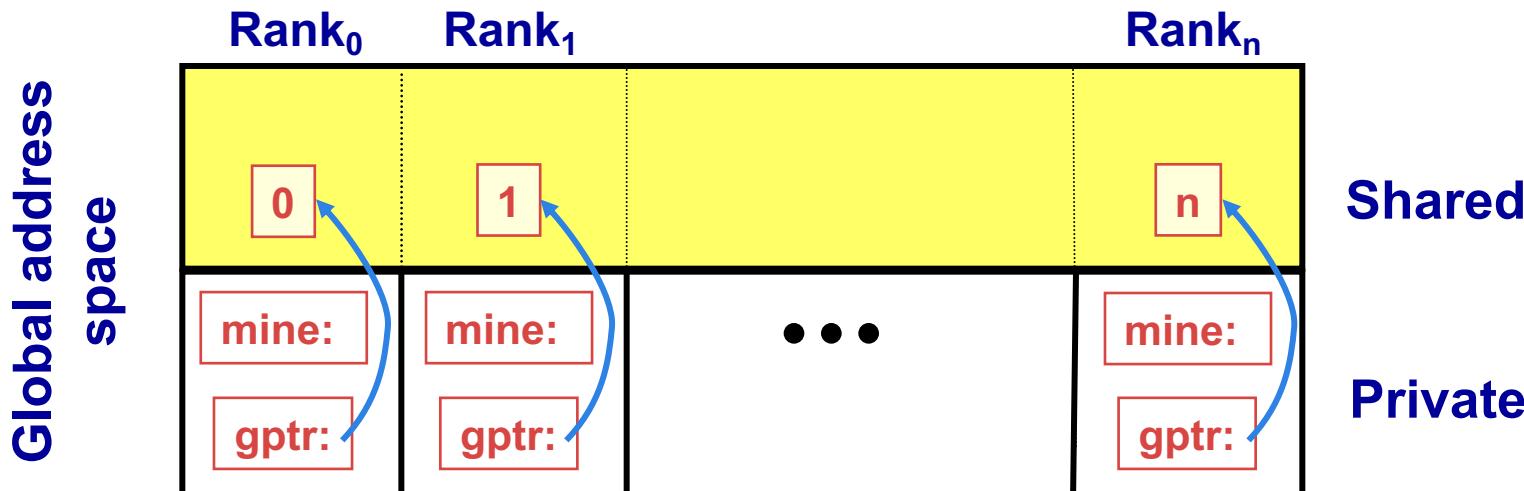
UPC++ allows full use  
of the C++ Standard  
Template Library

# Private vs. Shared Memory in UPC++

- Normal C++ variables and objects are allocated in the private memory space for each thread
- Allocate in shared space with `new_`
- Free with `delete_`
- There are also array versions

```
int mine;  
global_ptr<int> gptr = new_<int>(rank_me());
```

upcxx:: qualifier elided  
from here on out  
UPC++ names in green



# Broadcast in UPC++

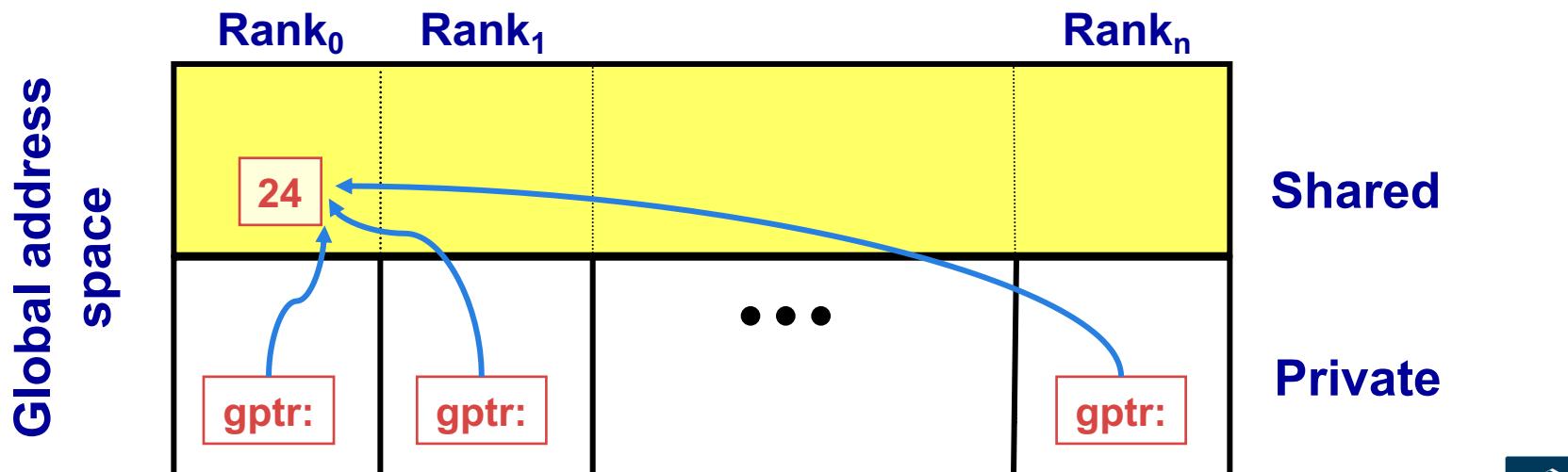
- To write an interesting program, we need to have global pointers refer to remote data
- One approach is to broadcast the pointer

```
global_ptr<int> gptr =  
    broadcast(new_<int>(24), 0).wait();
```

New int variable in  
shared space, set to 24

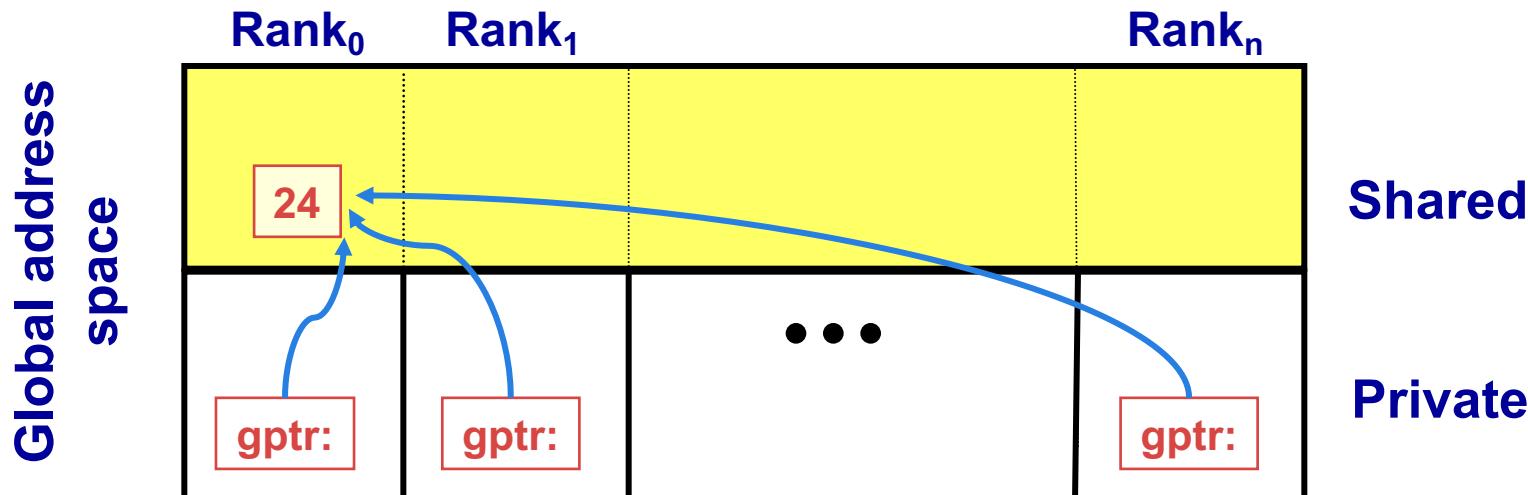
broadcast  
from rank 0

Will explain wait...



# Remote access: Put / Get

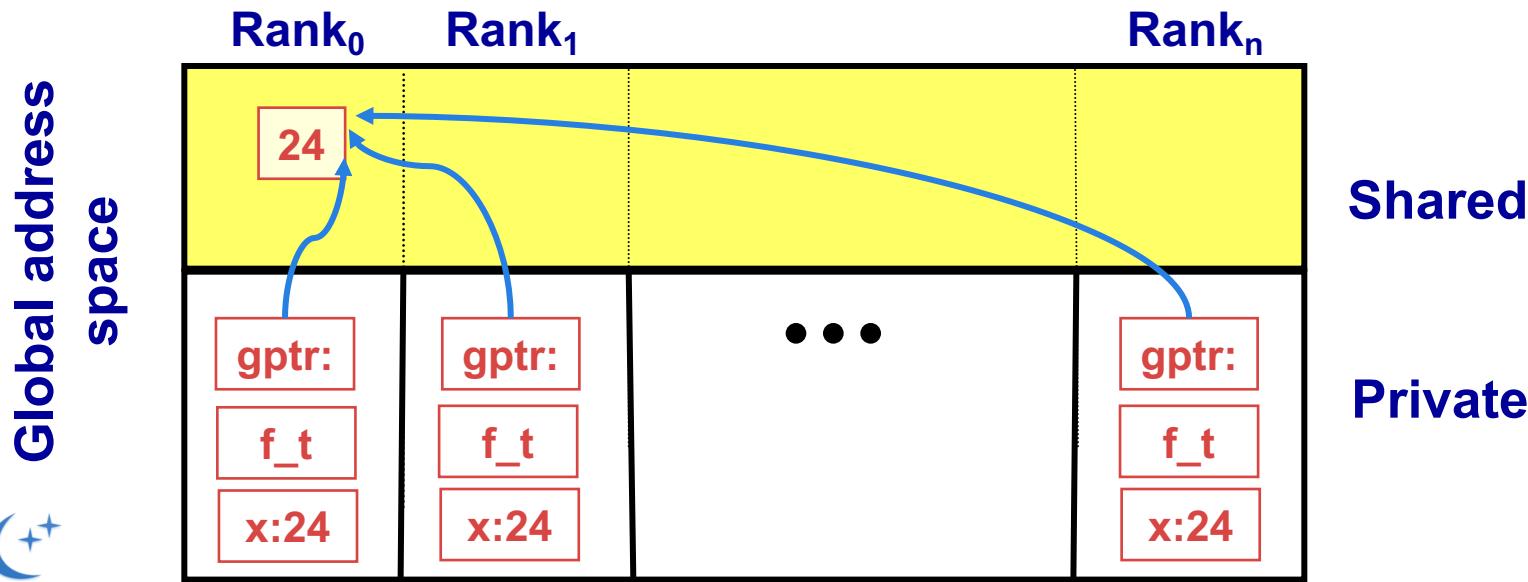
- We access shared memory from a remote rank
  - Remote get – read a variable on a remote rank
  - Remote put – write a variable on a remote rank
- Both done via a global pointer to the remote variable
- But will take microseconds at best!
  - And for most of that time, the processor is just waiting...



# Asynchronous remote operations

- Asynchronous execution used to hide remote latency
  - Asynchronous get: start reading, but how to tell if you're done?
  - Put the results into a special “box” called a future

```
future<int> fut_temp = rget(gptr);  
// ... Do something expensive  
int x = fut_temp.wait();  
// all ranks have x = 24
```



# Futures in general

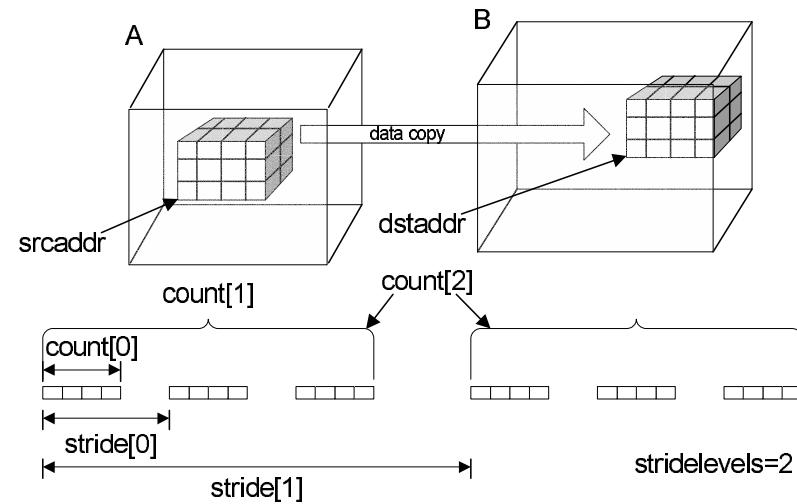
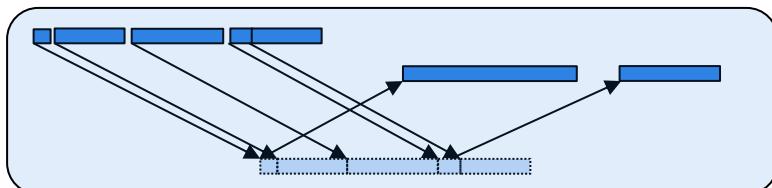
- A *future* holds a sequence of values and a state (ready / not ready)
- Waiting on the returned future lets user tailor degree of asynchrony they desire

```
future<T> f1 = rget(gptr1);      // asynchronous op
future<T> f2 = rget(gptr2);
// unrelated work...
bool ready = f1.ready();          // non-blocking poll
T t = f1.wait();                  // waits if not ready
```

- UPC++ has no *implicit* blocking
  - Except for synchronizing operations like “wait” others are implicitly nonblocking (asynchronous)

# One-Sided Communication in General

- In addition to these basic remote put/get operations
  - There is also a version that runs a handler at the destination *after* rput operation is visible at the target
  - And support for non-contiguous transfers



# UPC++ Synchronization

- UPC++ has two basic forms of barriers:
  - Barrier: block until all other threads arrive  
barrier() ;
  - Asynchronous barriers

```
future<> f =
    barrier_async(); // this thread is ready for barrier
// do computation unrelated to barrier
wait(f); // wait for others to be ready
```
- We will underline all blocking operations in these examples

# Pi in UPC++: Shared Memory Style

- Parallel computing of pi, but with a bug

```
int main(int argc, char **argv) {  
    init();                                divide work up evenly  
    int trials = atoi(argv[1]);  
    int my_trials = (trials+rank_me()) / rank_n();  
    global_ptr<int> hits =  
        broadcast(new<int>(0), 0).wait();  
    generator.seed(rank_me() * 17);  
    for (int i=0; i < my_trials; i++) {  
        int old_hits = rget(hits).wait();  
        rput(old_hits + hit(), hits).wait();  
    }  
    barrier();  
    if (rank_me() == 0)  
        cout << "PI estimated to " pointer that points locally  
              << 4.0 * (*hits.local()) / trials;  
    finalize();
```

What is the problem with this program?



# Atomics in UPC++

- Atomics are indivisible read-modify-write operations
- As if you put a lock around each operation, but may have hardware support (e.g., within the network interface)

Create “atomic domain” for shared ints that works with fetch-and-add and load operations.

```
atomic_domain<int> ad_int({atomic_op::load,
                             atomic_op::fetch_add});  
  
Locally  
“throw  
darts”  
  
int my_hits = 0;  
for (int i=0; i < my_trials; i++) my_hits += hit();  
ad_int.fetch_add(hits, my_hits,  
                  memory_order_relaxed).wait();  
  
Atomically  
update shared hits variable C++ memory order: relaxed says other memory  
accessed not constrained  
barrier();  
if (rank_me() == 0)  
    cout << "PI estimated to " << 4.0*ad_int.load(  
        hits, memory_order_relaxed).wait()  
        /trials;
```

Once a variable is used with  
atomics, always use atomics

# Remote Procedure Call

```
future<R> rpc(intrank_t r,  
                 F func, Args&&... args);
```

- Executes `func(args...)` on rank `r` and returns the result
- `R` is the return type of `func`
  - Empty future if `func` returns `void`
- There is also a ‘fire and forget’ version that returns no result
- Some restrictions apply to what UPC++ operations can be issued in an RPC: the *restricted context*
  - Limits on blocking operations from within an RPC

# Pi in UPC++: RPC

- RPC used to synchronize updates

`int hits = 0; RPC can refer to global variable`

```
int main(int argc, char **argv) {
    init();
    int trials = atoi(argv[1]);
    int my_trials = (trials+rank_me()) / rank_n();
    generator.seed(rank_me()*17);
    for (int i=0; i < my_trials; i++) {
        rpc(0, [](int hit) { hits += hit; },
            hit()).wait();
    }
    send update to rank 0
    barrier(); block on the update
    if (rank_me() == 0)
        cout << "PI estimated to " << 4.0*hits/trials;
    finalize(); Similar to atomics, but always runs on the remote
processor (not in network) and can be arbitrary
function (not just simple operation)
```

# UPC++ Collectives

- UPC++ has a small set of collectives (so far)
- They are all asynchronous
  - Broadcast (value, sender): get a value from one rank

```
template <typename T> future <T>
broadcast (T && value , intrank_t root);
```
  - Broadcast (buffer, count, sender): get a buffer of values from one rank

```
template <typename T> future <T>
broadcast (T * buffer, std::size_t count,
intrank_t sender);
```
  - Reduce (value, op): combine values across ranks

```
template <typename T, typename BinaryOp>
future <T> reduce_all (T && value, BinaryOp &&op);
```
- All of these take an optional “team” of ranks and have more general completion semantics (details in UPC++ guide)

# Pi in UPC++: Data Parallel Style w/ Collectives

- The previous version of Pi works, but is not scalable:
  - Updates are serialized on rank 0, ranks block on updates
- Use a reduction for better scalability:

```
// int hits; no global variables or shared memory
int main(int argc, char **argv) {
    ...
    for (int i=0; i < my_trials; i++)
        my_hits += hit();
    int hits = reduce_all(my_hits,
                          op_fast_add).wait();
    // barrier(); barrier implied by reduce +wait
    if (rank_me() == 0)
        cout << "PI: " << 4.0*hits/trials;
    finalize();
}
```



# Distributed Objects

- Any C++ type can be made into a *distributed object*
- One instance on every rank of a team    **Mesh is a 3D array**

```
dist_object<int> counter(0); // over world team
```

- Collectively allocate a set of shared counters
- Can access remote instances

```
future<int> f = fetch(counter, someRank);  
int x = f.wait();
```

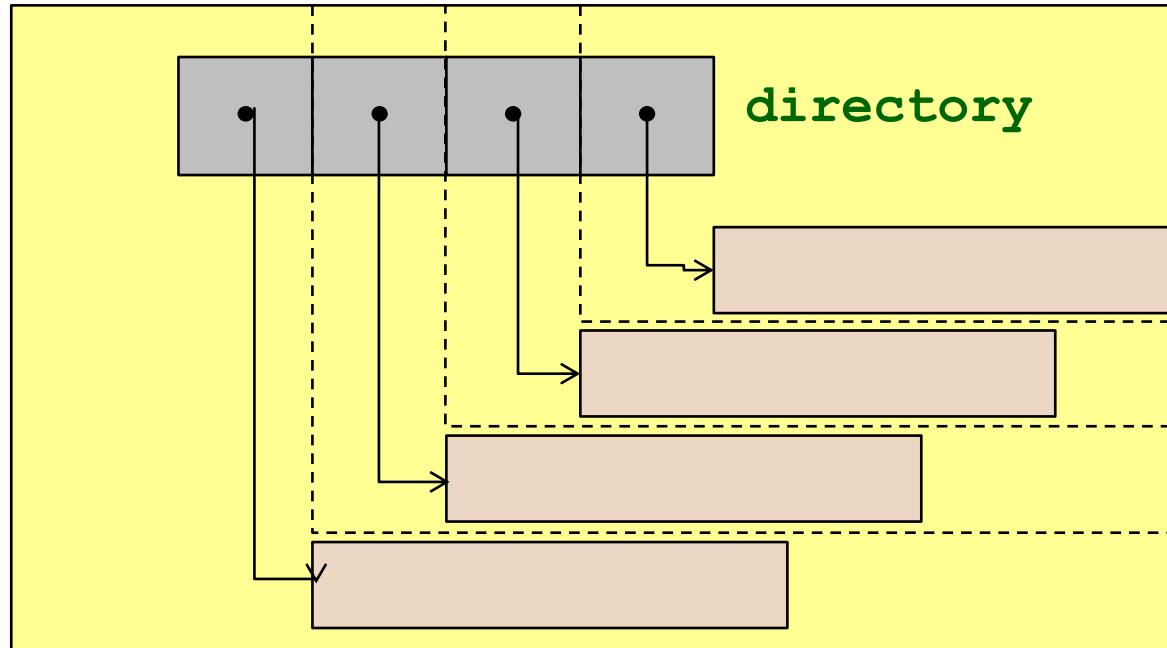
# Pi in UPC++: Distributed Object Version

- Alternative fix to the race condition
- Have each rank update a separate counter:
  - Do it in a distributed object, have one rank compute sum

```
int main(int argc, char **argv) {  
    ... declarations and initialization code omitted  
    dist_object<int> all_hits(0); all_hits  
distributed  
across all ranks  
    for (int i=0; i < my_trials; i++)  
        *all_hits += hit(); update element  
with local affinity  
    barrier(); collect each  
rank's  
    if (rank_me() == 0) {  
        for (int i=0; i < rank_n(); i++)  
            hits += fetch(all_hits, i).wait(); contribution  
        cout << "PI estimated to " << 4.0*hits/trials;  
    }  
    finalize();
```

# Distributed Arrays Directory Style

- A distributed object can be much more complicated than a set of ints
- Many UPC++ programs use these kinds of “directories” for distributed data structures



# Distributed Objects in Stencil Code

- Communication in 1D stencil (nearest-neighbor computation):



```
int main(int argc, char **argv) {  
    ... declarations and initialization code omitted  
    construct local grids  
    and distributed object
```

```
    global_ptr<double> my_grid =  
        new_array<int>(interior+2);  
    dist_object<global_ptr<double>> grids(my_grid);
```

```
    global_ptr<double> left =  
        fetch(grids, (rank_me() + rank_n() - 1) % rank_n()).wait();  
    global_ptr<double> right =  
        fetch(grids, (rank_me() + 1) % rank_n()).wait();
```

```
    for (int i=0; i < timesteps; i++) {  
        future<double> f1 = rget(left+interior);  
        future<double> f2 = rget(right+1);  
        ... wait on futures and do computation  
        get ghost cells
```

```
}
```

```
...
```

# Summary

- UPC++ is a PGAS library that supports lightweight communication over GASNet-EX
- Close to the metal performance, lean interface
  - Trade offs to reduce overheads and increase flexibility
    - Asynchronous and explicit communication
    - Reduced consistency guarantees
- Advanced features not covered in talk:
  - Promises, callbacks, remote atomics, progress, memory model, teams
- V1.0 release targeted for September 30, 2017
  - Will include programmer's guide



# Acknowledgements

- Early work with UPC++ involved Yili Zheng, Amir Kamil, Kathy Yelick, and others [IPDPS '14]
- Pagoda Project (GASNet-EX and UPC++): Scott B. Baden (PI), Paul Hargrove (co-PI), John Bachan, Dan Bonachea, Steven Hofmeyer, Khaled Ibrahim, Mathias Jacquelin, Amir Kamil, Brian van Straalen
- This research was supported by the Exascale Computing Project (17-SC-20-SC), funded by the U.S. Department of Energy
- UPC++ V1.0 draft specification available at  
<https://bitbucket.org/upcxx/upcxx>



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# PGAS Applications

UPC and old UPC++

Others (not shown) in Co-Array Fortran (Ocean model), Titanium (heart model) and Chapel

# Some of these started as CS267 projects

- Project ideas for this year:
- Contribute to the Berkeley Container Library
  - Data structure library (hash tables, bloom filters, matrices,... -- talk to Ben Brock)
- Compare UPC++ to something else
  - MPI, Ray, ...
- Write a more substantial algorithm in UPC++



# Application Challenge: Fast All-to-All

## Transpose in 3D FFT

- Three approaches:

- Chunk:**

- Wait for 2<sup>nd</sup> dim FFTs to finish
  - Minimize # messages

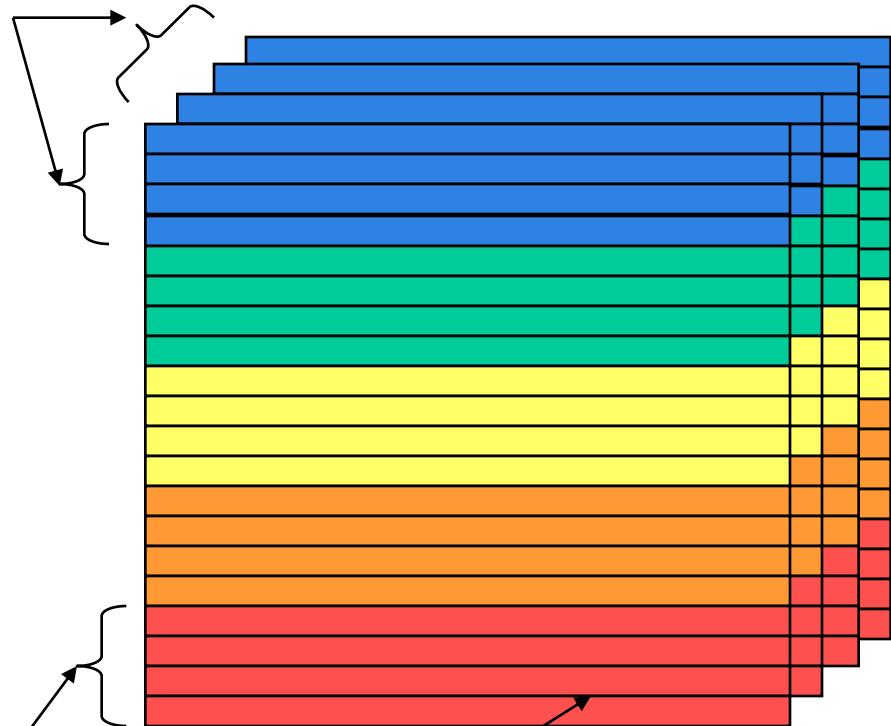
- Slab:**

- Wait for chunk of rows destined for 1 proc to finish
  - Overlap with computation

- Pencil:**

- Send each row as it completes
  - Maximize overlap and
  - Match natural layout

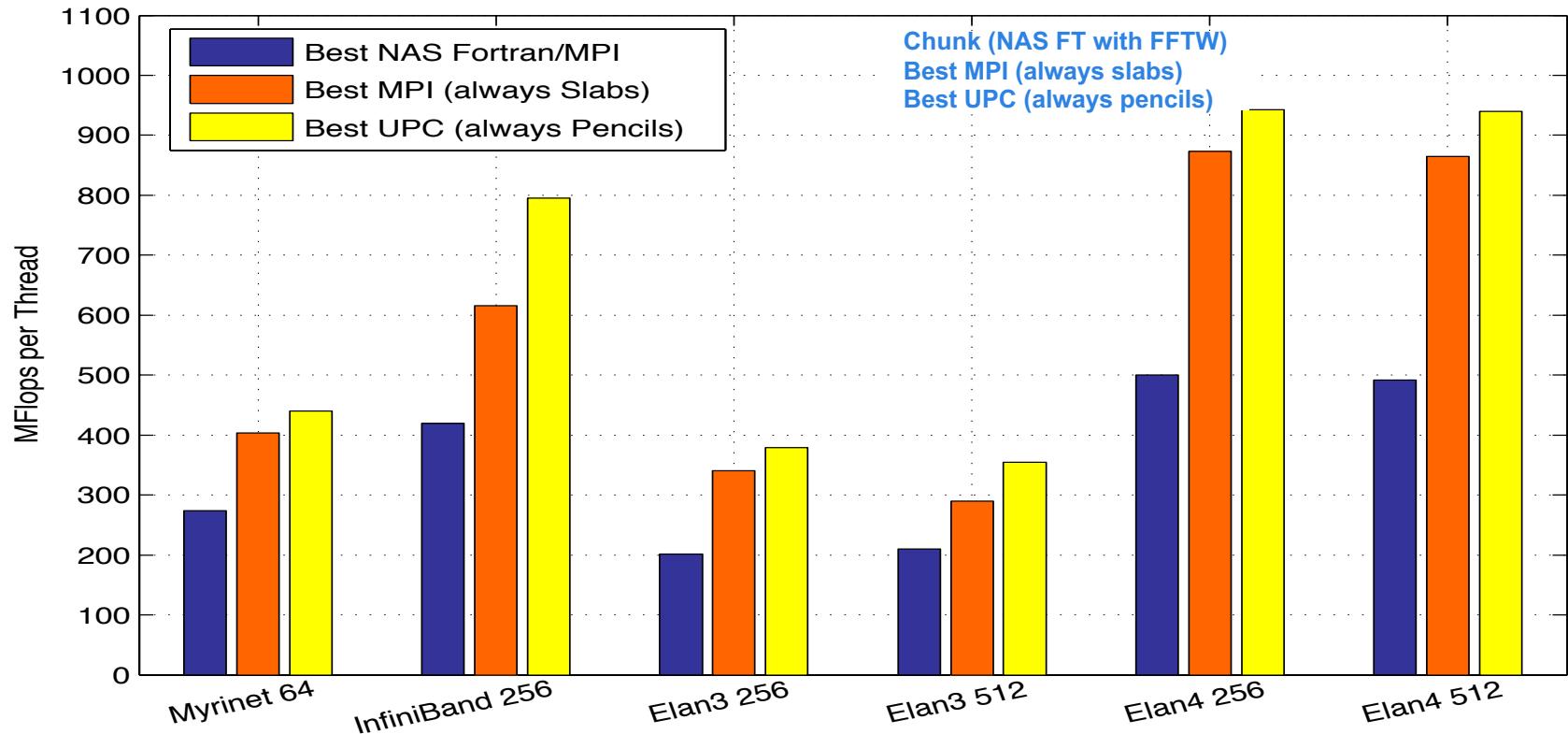
**chunk = all rows with same destination**



**pencil = 1 row**

**slab = all rows in a single plane with same destination**

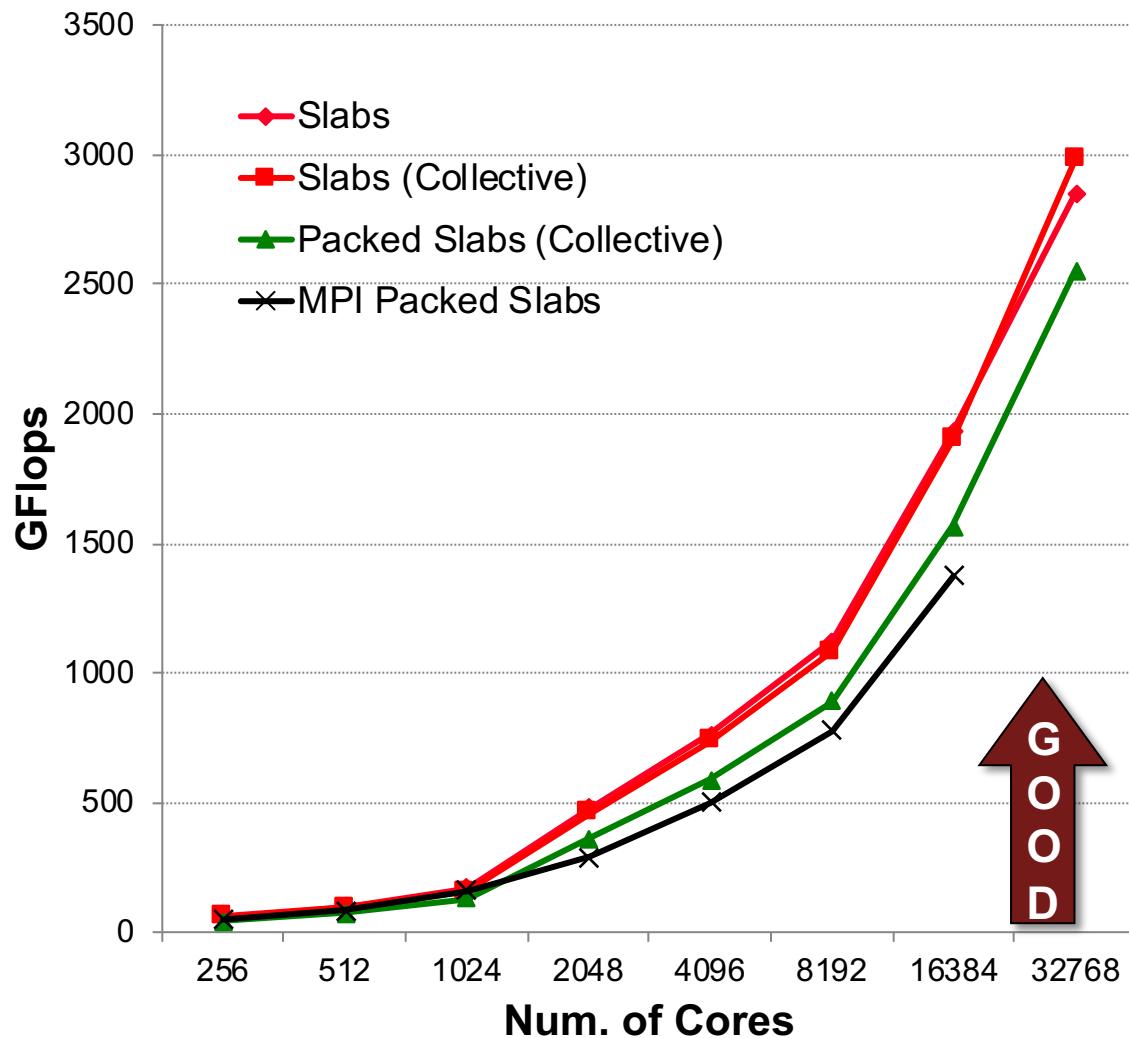
# Bisection Bandwidth



- Avoid congestion at node interface: allow all cores to communicate
- Avoid congestion inside global network: spread communication over longer time period (send early and often)

# FFT Performance on BlueGene/P (Mira)

- **UPC implementation outperforms MPI**
- **Both use highly optimized FFT library on each node**
- **UPC version avoids send/receive synchronization**
  - Lower overhead
  - Better overlap
  - Better bisection bandwidth



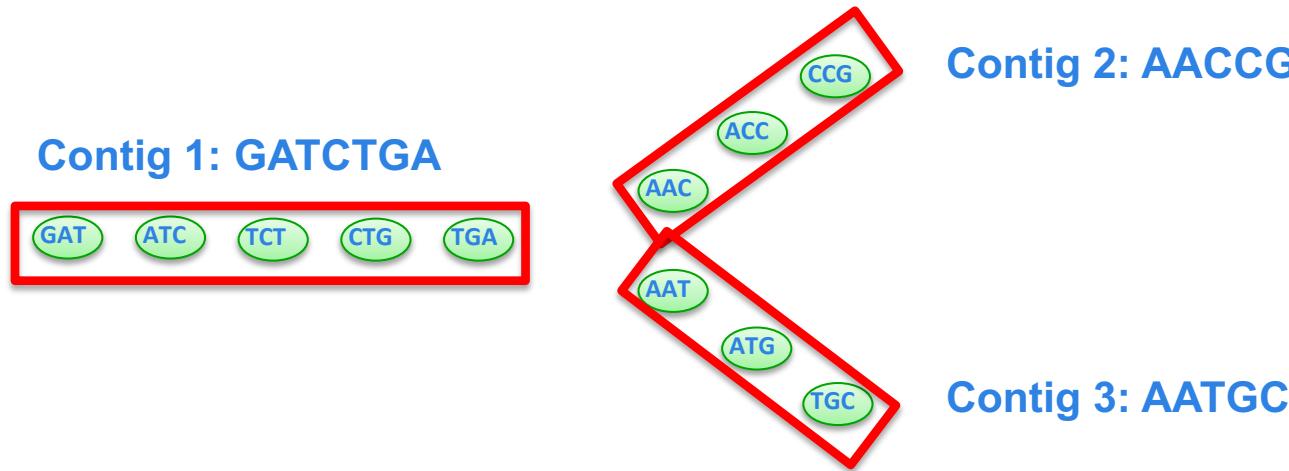
# *De novo* Genome Assembly

- DNA sequence consists of 4 bases: A/C/G/T
- Read: short fragment of DNA
- De novo assembly: Construct a genome (chromosomes) from a collection of reads



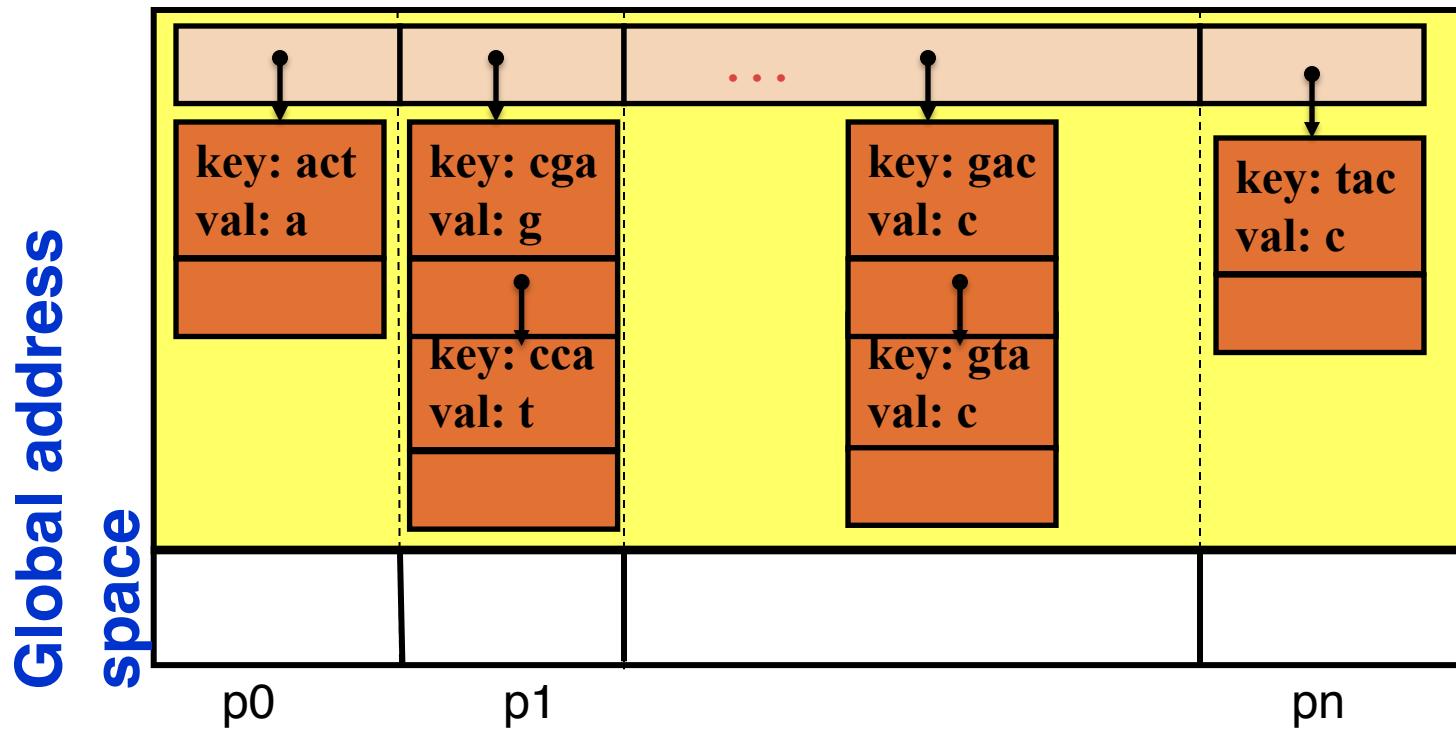
# PGAS in Genome Assembly

- Sequencers produce fragments called “reads”
- Chop them into overlap fixed-length fragments, “K-mers”
- Parallel DFS (from randomly selected K-mers) → “contigs”



- Hash tables used here (and in other assembly phases)
  - Different use cases, different implementations
- Some tricky synchronization to deal with conflicts

# Partitioned Global Address Space Programming



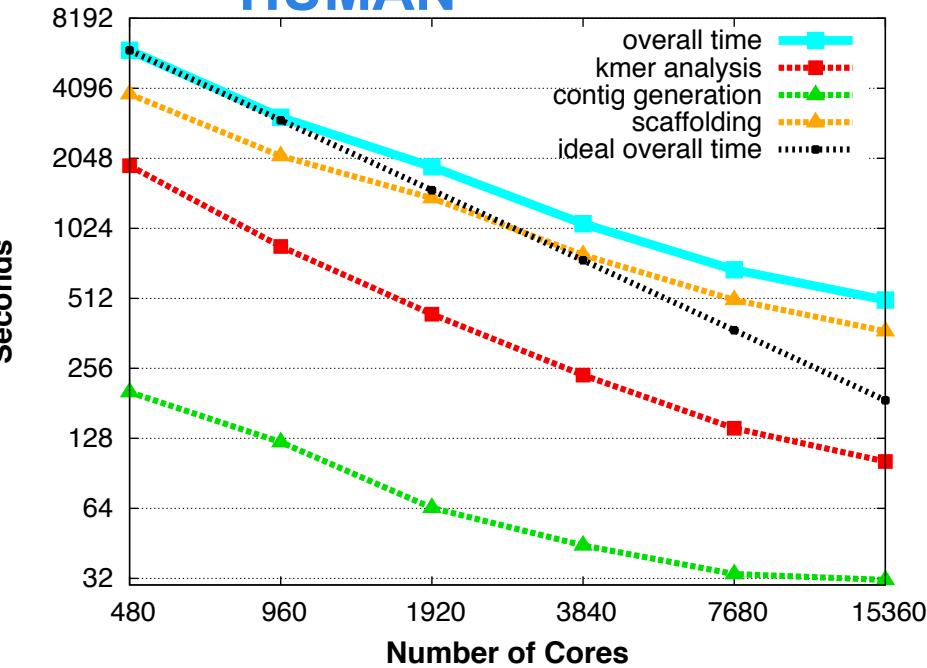
- Store the connections between read fragments (K-mers) in a hash table
- Allows for TB-PB size data sets

# HipMer (High Performance Meraculous) Assembly Pipeline

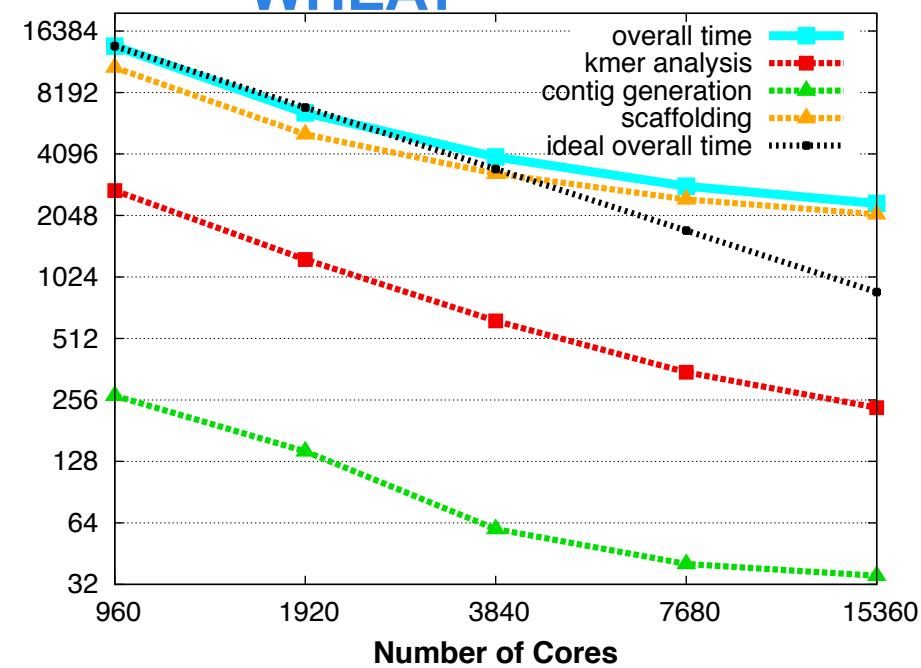
## Distributed Hash Tables in PGAS

- Remote Atomics, Dynamic Aggregation, Software Caching
- 13x Faster than MPI code (Ray) on 960 cores

HUMAN

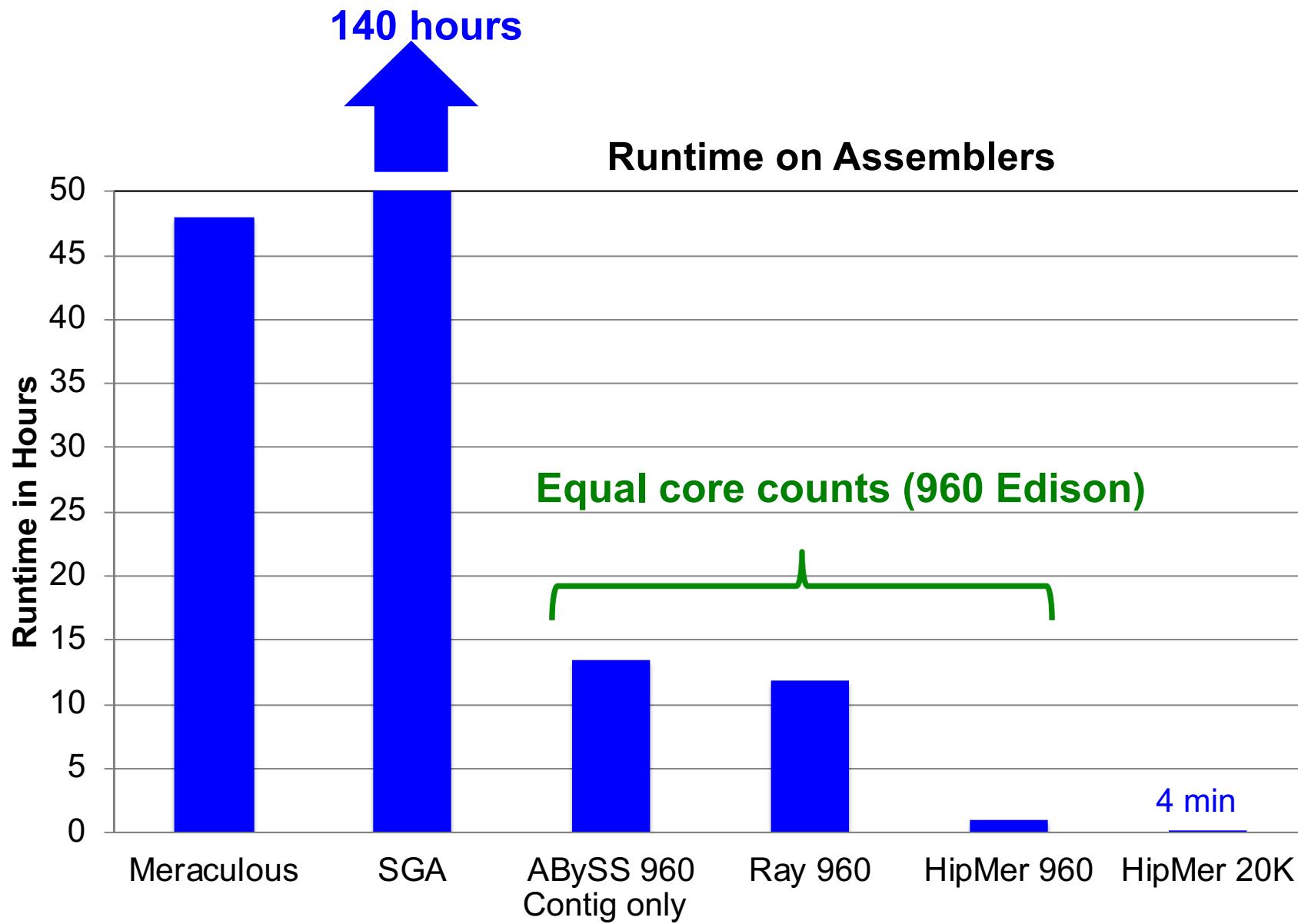


WHEAT



Evangelos Georganas, Aydin Buluc, Jarrod Chapman, Steven Hofmeyr, Chaitanya Aluru, Rob Egan, Lenny Oliker, Dan Rokhsar, and Kathy Yelick. HipMer: An Extreme-Scale De Novo Genome Assembler, SC'15

# Comparison to other Assemblers



# Science Impact: HipMer is transformative

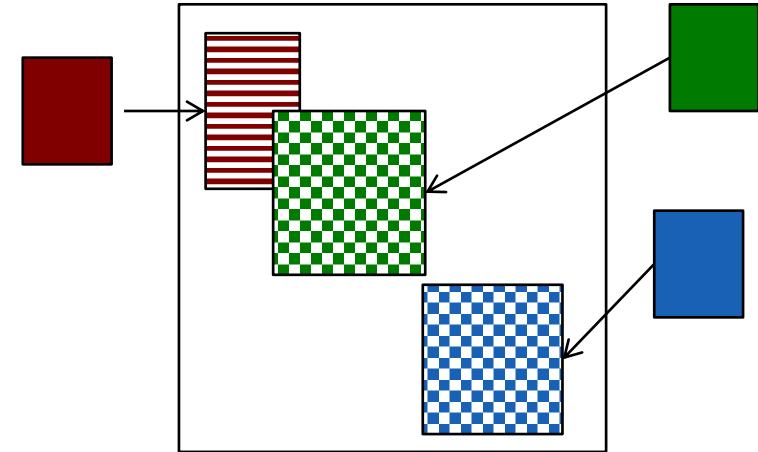
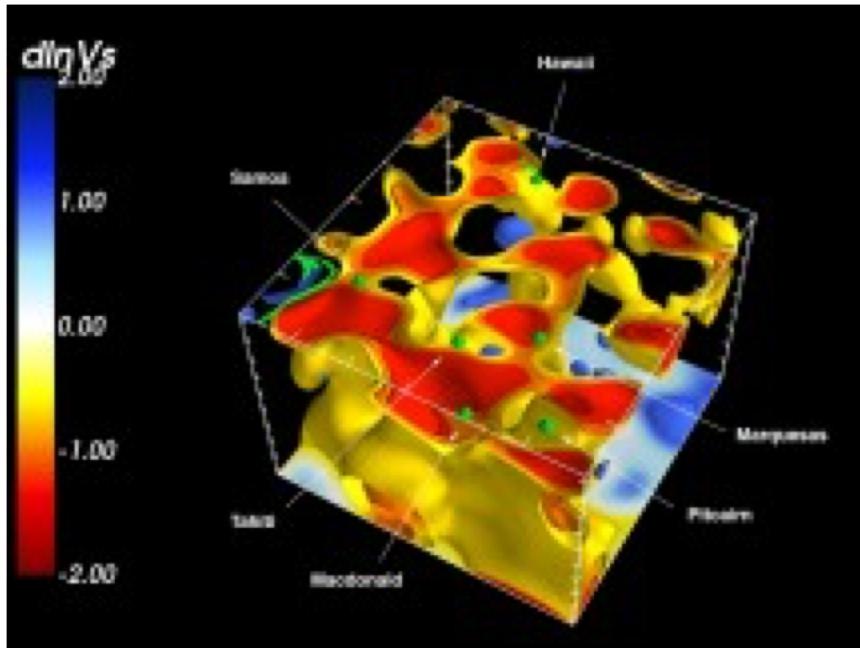
- Human genome (3Gbp) “de novo” assembled :
  - Meraculous: 48 hours
  - HipMer: 4 minutes (720x speedup relative to Meraculous)
- Wheat genome (17 Gbp) “de novo” assembled (2014):
  - Meraculous (did not run):
  - HipMer: 39 minutes; 15K cores (first all-in-one assembly)
- Pine genome (20 Gbp) “de novo” assembled (2014) :
  - Maserca : 3 months; 1 TB RAM
- Wetland metagenome (1.25 Tbp) analysis (2015):
  - Meraculous (projected): 15 TB of memory
  - HipMER: Strong scaling to over 100K cores (contig gen only)

Makes unsolvable problems solvable!



# Application Challenge: Data Fusion in UPC++

- Seismic modeling for energy applications “fuses” observational data into simulation
- With UPC++ “matrix assembly” can solve larger problems



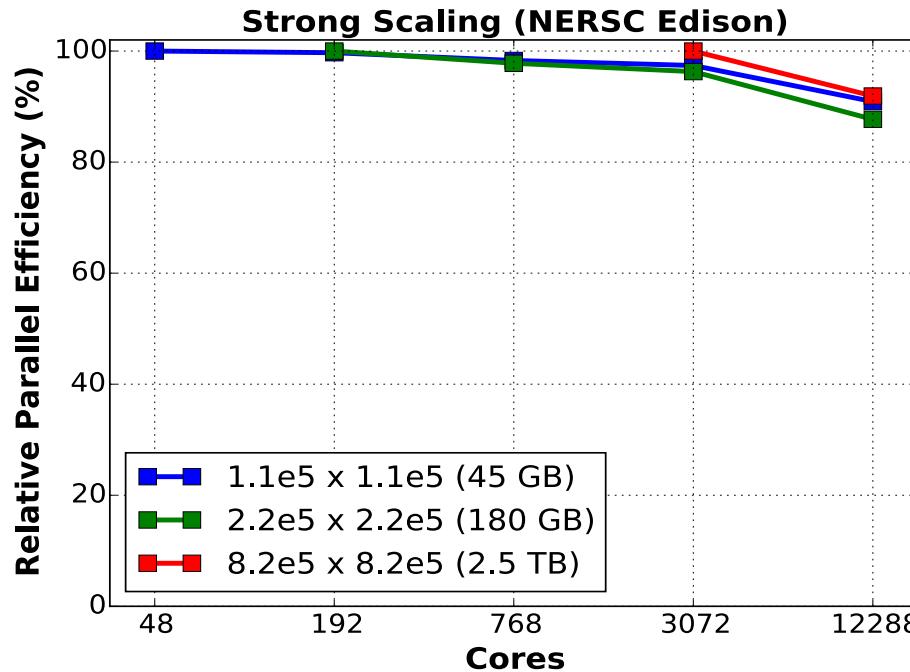
First ever sharp, three-dimensional scan of Earth’s interior that conclusively connects plumes of hot rock rising through the mantle with surface hotspots that generate volcanic island chains like Hawaii, Samoa and Iceland.



French and Romanowicz use code with UPC++ phase to compute *first* ever whole-mantle global tomographic model using numerical seismic wavefield computations (F & R, 2014, GJI, extending F et al., 2013, Science<sup>54</sup>).



# Application Challenge: Data Fusion in UPC++



## Distributed Matrix Assembly

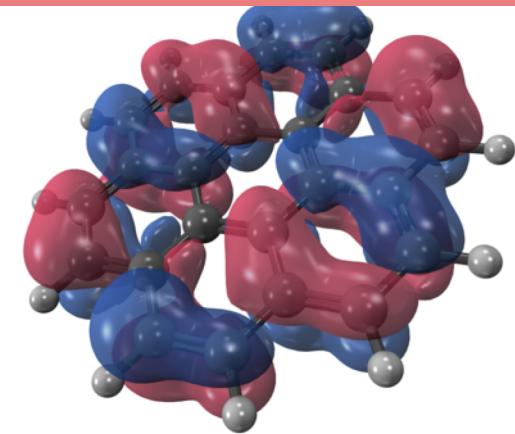
- Remote asyncs with user-controlled resource management
- Remote memory allocation
- Team idea to divide threads into injectors / updaters
- 6x faster than MPI 3.0 on 1K nodes
- Improving UPC++ team support

# Load Balancing and Irregular Matrix Transpose

- Hartree Fock example (e.g., in NWChem)

Increase scalability!

- Inherent load imbalance
- UPC++
  - Work stealing and fast atomics
  - Distributed array: easy and fast transpose
- Impact
  - *20% faster than the best existing solution (GTFock with Global Arrays)*



	0	1	2	3
Distributed Array	4	5	6	7
Local Array	8	9	10	11
	12	13	14	15

Local Array

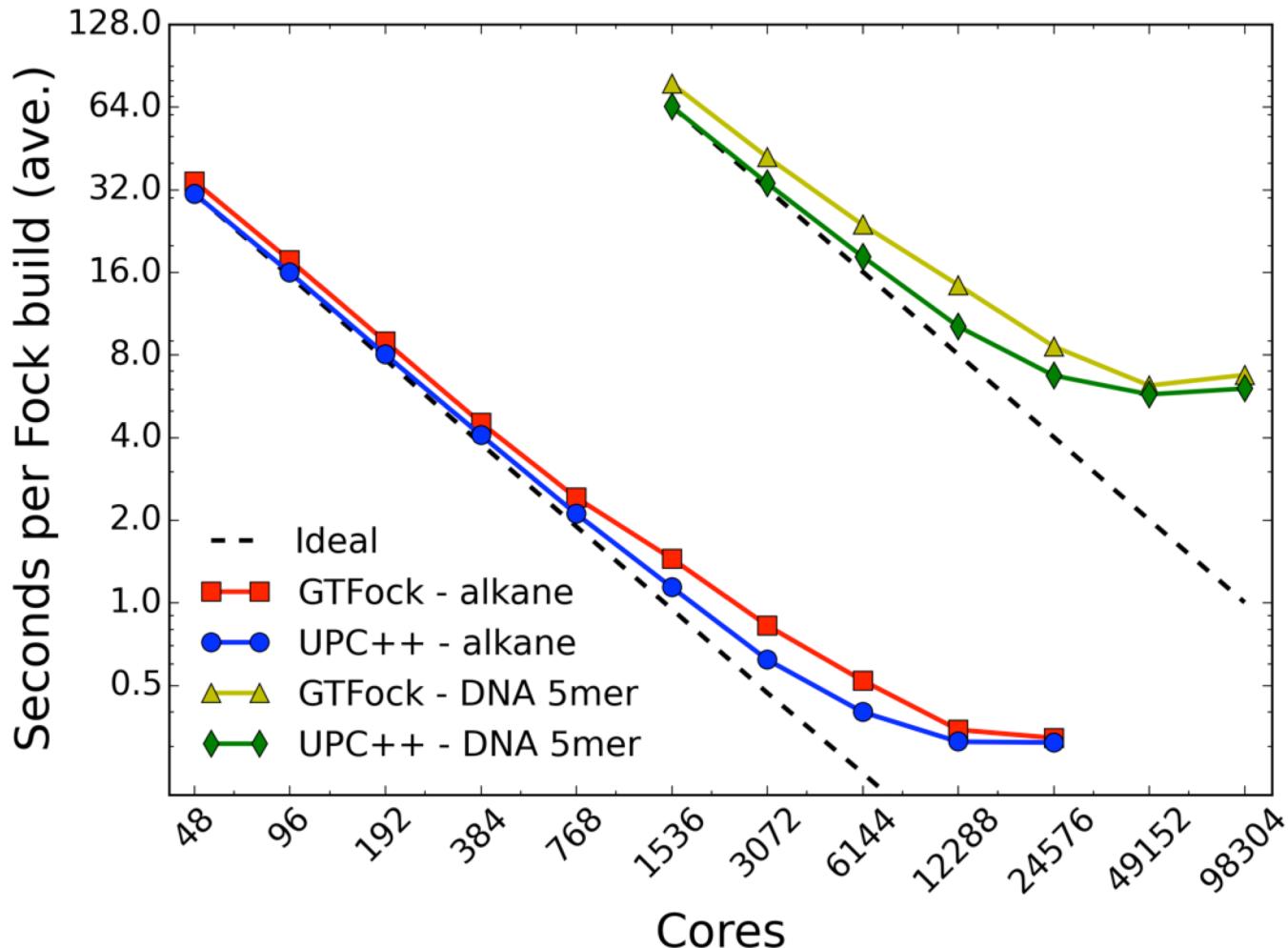


update

David Ozog , Amir Kamil , Yili Zheng, Paul Hargrove , Jeff R. Hammond, Allen Malony, Wibe de Jong, Katherine Yelick



# Hartree Fock Code in UPC++



**Strong Scaling of UPC++ HF Compared to GTFock with Global Arrays on NERSC Edison (Cray XC30)**

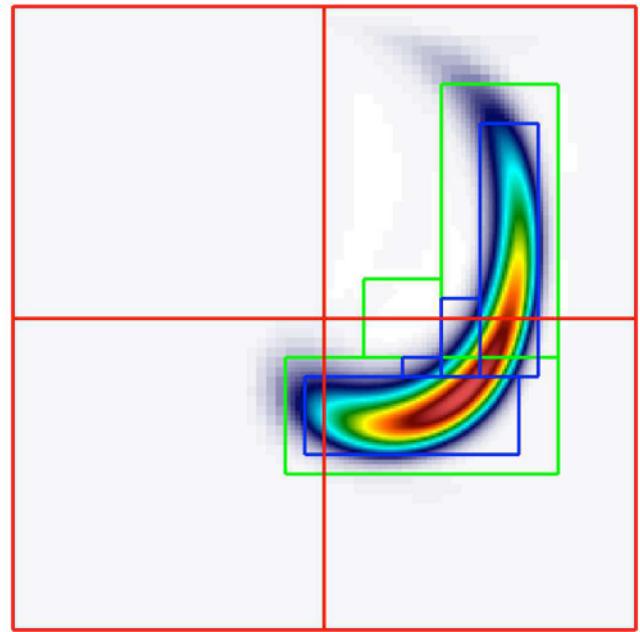
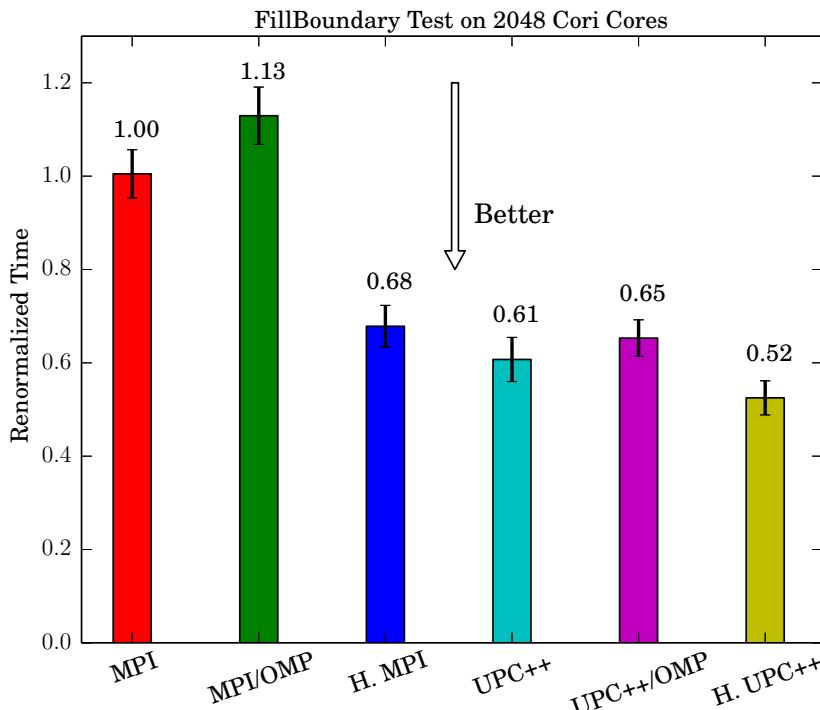


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# UPC++ Communication Speeds up AMR

- Adaptive Mesh Refinement on Block-Structured Meshes
  - Used in ice sheet modeling, climate, subsurface (fracking),

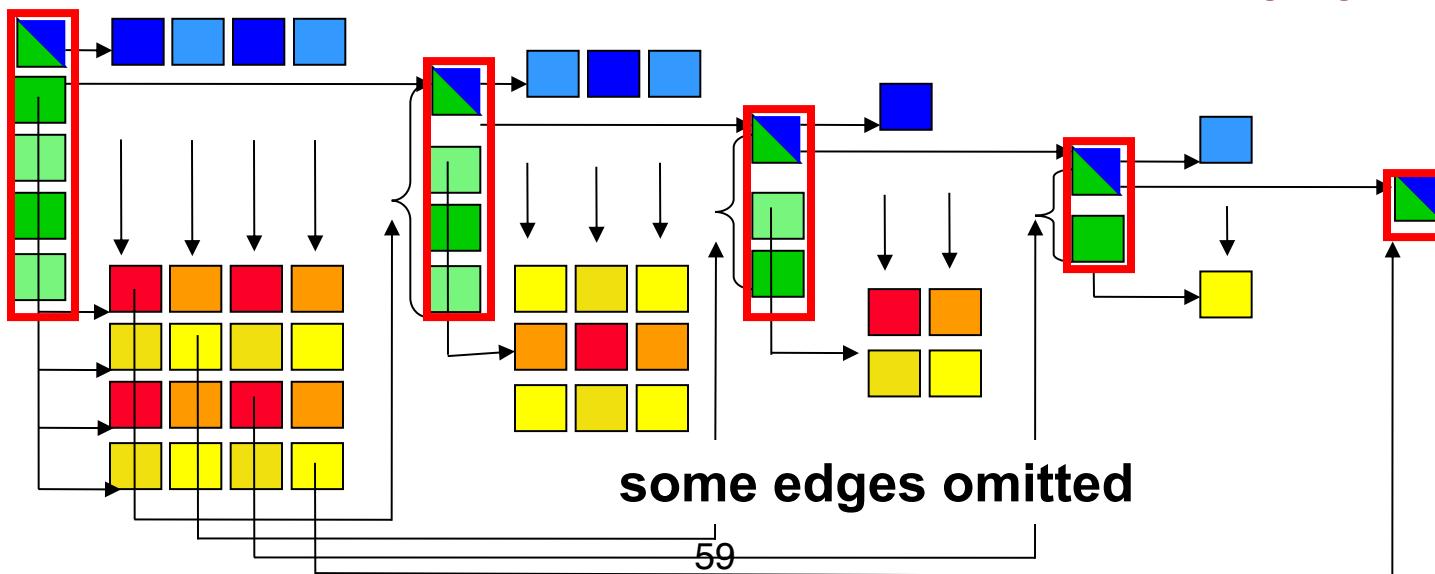


- Hierarchical UPC++ (distributed / shared style)
- UPC++ plus UPC++ is 2x faster than MPI plus OpenMP
  - MPI + MPI also does well

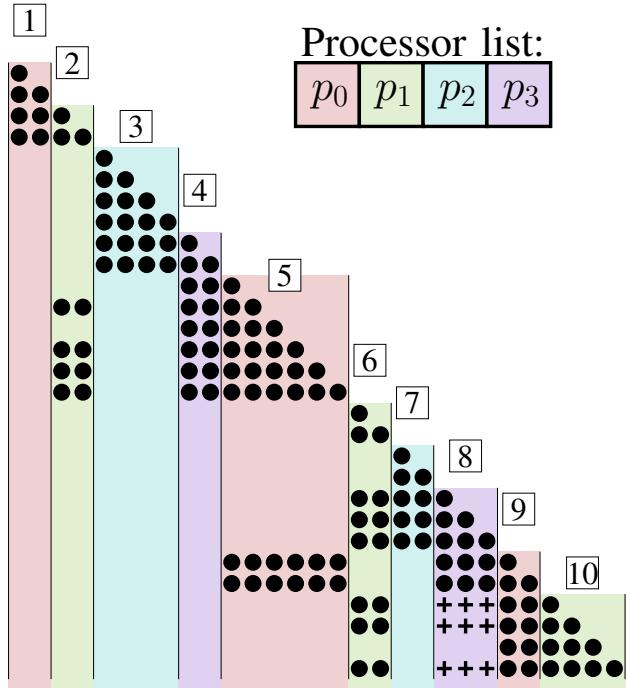
# Beyond Put/Get: Event-Driven Execution

- DAG Scheduling in a distributed (partitioned) memory context
- Assignment of work is static; schedule is dynamic
- Ordering needs to be imposed on the schedule
  - Critical path operation: Panel Factorization
- General issue: dynamic scheduling in partitioned memory
  - Can deadlock in memory allocation
  - “memory constrained” lookahead

**Uses a Berkeley extension to UPC to remotely synchronize**



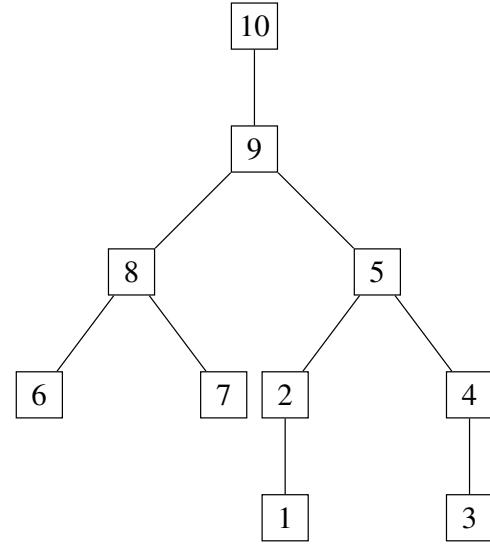
# symPACK: Sparse Cholesky



(a) Structure of Cholesky factor  $L$  (

Processor list:

$p_0$   $p_1$   $p_2$   $p_3$



) Supernodal elimination tree of matrix  $A$

- Sparse Cholesky using fan-bot algorithm in UPC++
    - Uses asynchronous tasks with dependencies
- Matthias Jacquelin, Yili Zheng, Esmond Ng, Katherine Yelick

# symPACK: Sparse Cholesky

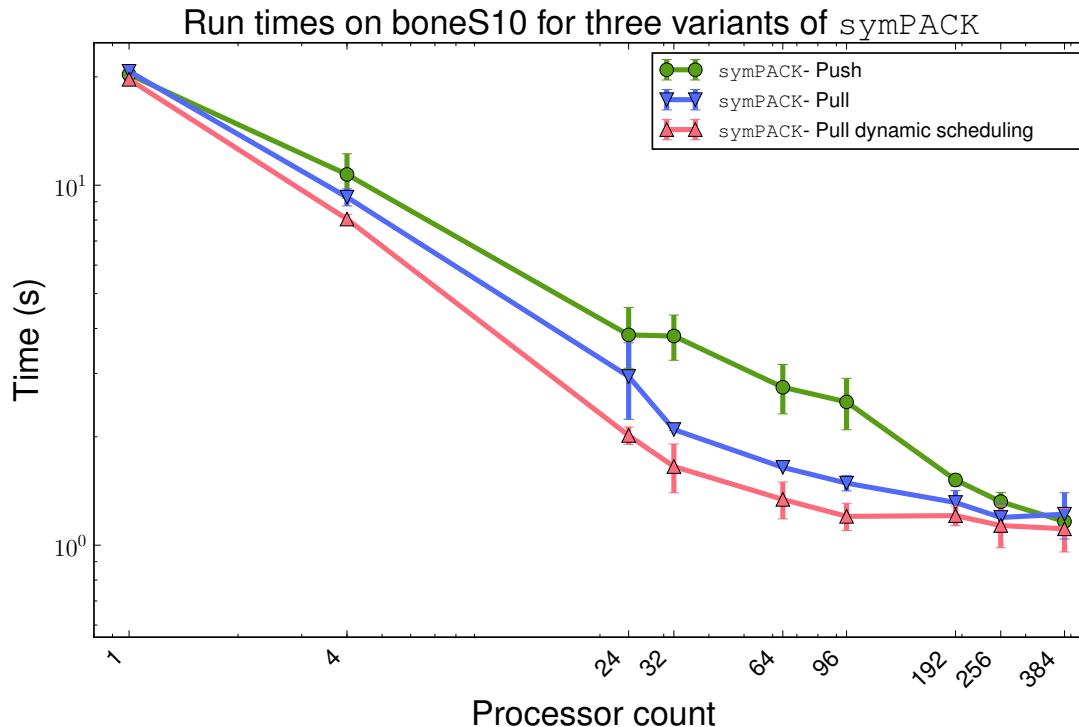


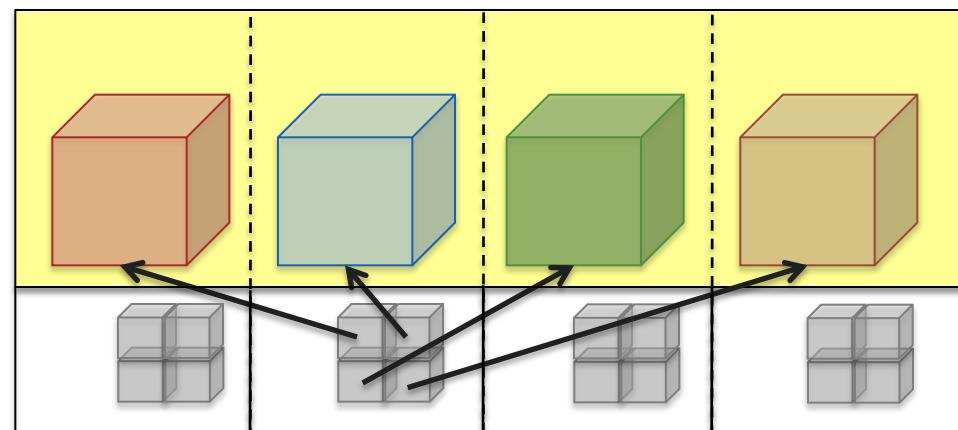
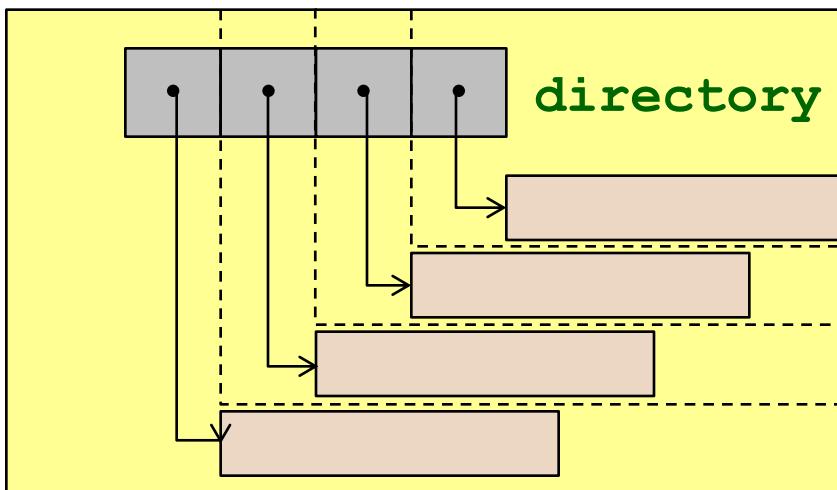
Figure 7: Impact of communication strategy and scheduling on symPACK performance

- Scalability of symPACK on Cray XC30 (Edison)
  - Comparable or better than best solvers (evaluation in progress)
  - Notoriously hard parallelism problem

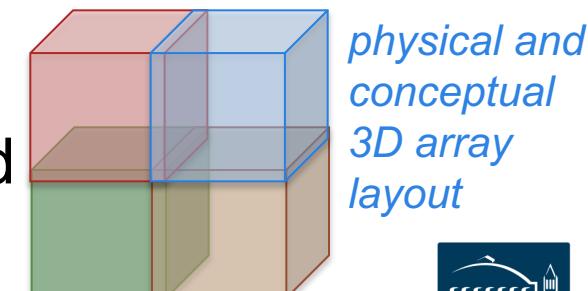
# Common Pattern for Distributed Data Structures

- Many UPC programs avoid the UPC style arrays in favor of directories of objects

```
typedef shared [] double *sdblptr;  
shared sdblptr directory[THREADS];  
directory[i]=upc_alloc(local_size*sizeof(double));
```



- These are also more general:
  - Multidimensional, unevenly distributed
  - Ghost regions around blocks



# Summary: PGAS for Irregular Applications

- Lower overhead of communication makes PGAS useful for latency-sensitive problems or bisection bandwidth problems
- Specific application characteristics that benefit:
  - Fine-grained updates (Genomics HashTable construction)
  - Latency-sensitive algorithms (Genomics DFS)
  - Distributed task graph (Cholesky)
  - Work stealing (Hartree Fock)
  - Irregular matrix assembly / transpose (Seismic, HF)
  - Medium-grained messages (AMR)
  - All-to-all communication (FFT)
- There are also benefits of thinking algorithmically in this model: parallelize things that are otherwise hard to imagine



# Summary: PGAS for Modern HPC Systems

- The lower overhead of communication is also important given current machine trends
  - **Many lightweight cores** per node (do not want a hefty serial communication software stack to run on them)
  - **RDMA mechanisms** between nodes (decouple synchronization from data transfer)
  - **GAS on chip**: direct load/store on chip without full cache coherence across chip
  - **Hierarchical machines**: fits both shared and distributed memory, but supports hierarchical algorithms
  - **New models of memory**: High Bandwidth Memory on chip or NVRAM above disk

# Using UPC++ at NERSC

Switch to the gnu C++ compiler

```
module switch PrgEnv-intel PrgEnv-gnu
```

Load the UPC++ module via

```
module load upcxx
```



Compile code with the upcc

```
upcxx=<upcxx-install-path>/bin/upcxx-meta"  
g++ --std=c++11 hello-world.cpp $( $upcxx PPFLAGS ) \  
    $( $upcxx LDFLAGS ) $( $upcxx LIBFLAGS )
```

Or see the Makefile example...

# Installing Berkeley UPC++, UPC, and GASNet

Available on Mac OSX, Linux, Infiniband clusters, Ethernet clusters, and most HPC systems

- UPC++ Open source with BSD license  
<https://bitbucket.org/berkeleylab/upcxx/>  
Has specification, user guide, codes, and instructions
- GASNet communication  
<https://gasnet.lbl.gov>

Not used in this course (this year)

- Cray UPC (installed at NERSC)  
<http://docs.cray.com/books/S-2179-50/>
- Berkeley UPC  
<http://upc.lbl.gov>

