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Programming Challenges and Solutions



Message Passing Programming

Divide up domain in pieces Each compute one piece Exchange (send/receive) data

Global Address Space Programming

Each start computing Grab whatever you need whenever

PVM, MPI, and many libraries

Global Address Space Languages and Libraries

~10% of NERSC apps use some kind of PGAS-like model



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
- In shared memory
 - -Lock each bucket

- Distributed memory: the array is huge and spread out
 - -Each processor has a substream and sends +1 to the appropriate processor... and that processor "receives"





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







PGAS Languages







UPC++ Features



UPC++: PGAS with "Mixins"

UPC++ uses templates (no compiler needed)

shared_var<int> s;
global_ptr<LLNode> g;
shared_array<int> sa(8);

- Default execution model is SPMD, but
- Remote methods, async
 async(place) (Function f, T1 arg1,...);
 async_wait(); // other side does poll();



 Research in teams for hierarchical algorithms and machines

teamsplit (team) { ... }

Interoperability is key; UPC++ can be use with OpenMP or MPI





Why Should You Care about PGAS?



Random Access to Large Memory

Meraculous Genome Assembly Pipeline



Contig generation step:

upC

- Human: 44 hours to 20 secs
- Wheat: "doesn't run" to 32 secs

Grand Challenge: Metagenomes



Graph as Distributed Hash Table

- Remote Atomics
- Dynamic Aggregation
- Software Caching
- Fast I/O (HDF5)
- Bloom filters, locality-aware hashing,...



~20% of Edison @ NERSC can assemble all human genomes produced worldwide in 2015



UPC++ Execution Model

UPC++ Basics

- UPC++ reserves all names that start with UPCXX or upcxx, or that are in the upcxx namespace
- Include "upcxx.h" for using UPC++
- Init and finalize the runtime

int upcxx::init(&argc, &argv);
int upcxx::finalize();

 Number of processes in the parallel job and my ID uint32_t upcxx::ranks(); // THREADS in UPC uint32_t upcxx::myrank(); // MYTHREAD in UPC

Tip: Add "using namespace upcxx;" to save typing "upcxx::"





Hello World in UPC++

- Any legal C/C++ program is also a legal UPC++ program
- If you compile and run it with P processes, it will run P copies of the program, also known as SPMD

```
#include <upcxx.h>
#include <iostream>
```

```
using namespace upcxx; // save typing "upcxx::"
```



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 = $\frac{1}{4} * \pi r^2 = \frac{\pi}{4}$

- Randomly throw darts at x,y positions
- If $x^2 + y^2 < 1$, then point is inside circle
- Compute ratio:
 - -# points inside / # points total
 - $-\pi = 4$ *ratio







Pi in UPC++ (ported from the UPC version)

 Independent estimates of pi: main(int argc, char **argv) int i, hits, trials = 0; Each thread gets its own copy of these variables double pi; Each thread can use if (argc != 2) trials = 1000000; input arguments else trials = atoi(argv[1]); Initialize random in srand(myrank()*17); math library for (i=0; i < trials; i++) hits += hit();</pre> pi = 4.0*hits/trials; printf("PI estimated to %f.", pi);

Each thread calls "hit" separately





Helper Code for Pi in UPC++ (same as UPC)

• Required includes:

```
#include <stdio.h>
#include <math.h>
#include <upcxx.h> // #include <upc.h> for UPC
```

• Function to throw dart and calculate where it hits:

```
int hit() {
    int const rand_max = 0xFFFFFF;
    double x = ((double) rand()) / RAND_MAX;
    double y = ((double) rand()) / RAND_MAX;
    if ((x*x + y*y) <= 1.0) {
        return(1);
    } else {
        return(0);
    }
}</pre>
```





Shared vs. Private Variables

Private vs. Shared Variables in UPC++

- Normal C++ variables and objects are allocated in the private memory space for each rank.
- Shared variables are allocated only once, with thread 0 shared_var<int> ours; // use sparingly: performance int mine;
- Shared variables may not have dynamic lifetime: may not occur in a function definition, except as static. Why?



Declaration:

```
shared_var<T> ours;
```

Explicit read and write with member functions get and put
T ours.get();
ours.put(const T& val);

Implicit read and write a shared variable in an expression

- Type conversion operator "T()" is overloaded to call get int mine = ours; // C++ compiler generates an implicit type conversion from shared_var<T> to T
- Assignment operator "=" is overloaded to call put

ours = 5;

 Compound operators such as "+=" and "-=" involve both a read and a write. Note that these are not atomic operations.



Pi in UPC++: Shared Memory Style



up(+

Pi in UPC: Shared Memory Style







Declaration:

shared_array<Type> sa;

Initialization (should be called collectively):
 sa.init(size_t array_size, sizt_t blk_size);

Finalization (should be called collectively)
sa.finalize();

Accessing Arrays elements:

sa[index] = ...; ... = sa[index]; cout << sa[index];</pre>





Shared Arrays Are Cyclic By Default

- Shared scalars always live in thread 0
- Shared arrays are spread over the ranks
- Shared array elements are spread across the processes shared_array<int> x, y, z;

 - y.init(3*ranks()); /* 3 elements per process */
 - /* 2 or 3 elements per process */
- In the pictures below, assume ranks() = 4

z.init(3*3);



Pi in UPC: Shared Array Version

- Alternative fix to the race condition
- Have each thread update a separate counter:
 - -But do it in a shared array
 - -Have one thread compute sum

shared_array<int> all_hits;

main(int argc, char **argv) {

all_hits.init(ranks());

for (i=0; i < my_trials; i++)</pre>

all_hits[myrank()] += hit();

barrier();

all_hits is shared by all processors, just as hits was

update element with local affinity

if (myrank() == 0) {
 for (i=0; i < ranks(); i++) hits += all_hits[i];
 printf("PI estimated to %f.", 4.0*hits/trials);
}</pre>



Asynchronous Task Execution

UPC++ Async

• C++ 11 async function
std::future<T> handle
= std::async(Function&& f, Args&&... args);
handle.wait();

```
    UPC++ async function
        // Remote Procedure Call
        upcxx::async(rank)(Function f, T1 arg1, T2 arg2,...);
        upcxx::async_wait();
```

```
// Explicit task synchronization
upcxx::event e;
upcxx::async(place, &e)(Function f, T1 arg1, ...);
e.wait();
```



up(+

#include <upcxx.h>

```
void print_num(int num)
  printf("myid %u, arg: %d\n", MYTHREAD, num);
}
int main(int argc, char **argv)
{
 for (int i = 0; i < upcxx::ranks(); i++) {</pre>
    upcxx::async(i)(print_num, 123);
  }
  upcxx::async_wait(); // wait for all remote tasks to complete
  return 0;
}
```





using namespace upcxx;

1003

num:

```
// Rank 0 spawns async tasks
for (int i = 0; i < ranks(); i++) {</pre>
  // spawn a task expressed by a lambda function
  async(i)([] (int num)
              { printf("num: %d\n", num); },
              1000+i); // argument to the \lambda function
}
async_wait(); // wait for all tasks to finish
       mpirun –n 4 ./test_async
       Output:
              1000
       num:
       num: 1001
       num: 1002
```









Example: Building A Task Graph

```
using namespace upcxx;
                                        t2
                                   t1
event e1, e2, e3;
                                      е
                                      1
async(P1, &e1)(task1);
async(P2, &e1)(task2);
                                        t3
async_after(P3, &e1, &e2)(task3);
async(P4, &e2)(task4);
async_after(P5, &e2, &e3)(task5);
async_after(P6, &e2, &e3)(task6);
                                         t5
async_wait(); // all tasks will be done
```





Progress Function for Async Tasks

 Each UPC++ rank decides when to execute incoming tasks and send outgoing tasks by polling the task queue:

int advance(int max_in, int max_out)

- max_in maximum number of incoming tasks to be processed before returning
- max_out maximum number of outgoing tasks to be processed before returning
- Support different progress models, for example:
 - Call advance() from the default thread
 - Create a dedicated progress thread for polling
- Important Progress Properties
 - Blocking functions in UPC++ call advance() internally to guarantee progress. These only include: async_wait(), barrier(), event.wait(), finish and finalize().
 - Other UPC++ functions are non-blocking





Dynamic Memory Management and Bulk Data Transfer

Dynamic Global Memory Management

- Global address space pointers (pointer-to-shared) global_ptr<Type> ptr;

void deallocate(global_ptr<T> ptr);

Example: allocate space for 512 integers on rank 2
global_ptr<int> p = allocate<int>(2, 512);

Remote memory allocation is not available in MPI-3, UPC or SHMEM.





More on Global Pointers

- Global pointers examples: global_ptr<int> p1; global_ptr<void> p2; global_ptr<void *> p3;
- Query the location (owner) of the data Uint32_t where()
- Get the local pointer (virtual memory address)
 T* raw_ptr()
- Pointer arithmetic is the same as that for local pointers
 There is no phase field in the global pointer
- Can dereference a pointer to read from or write to the global location

*ptr or ptr[i]





One-Sided Data Transfer Functions

```
// Copy count elements of T from src to dst
upcxx::copy<T>(global_ptr<T> src,
            global_ptr<T> dst,
            size_t count);
```

```
// Implicit non-blocking copy
upcxx::async_copy<T>(global_ptr<T> src,
            global_ptr<T> dst,
            size_t count);
```

```
// Synchronize all previous asyncs
upcxx::async_wait();
```

Similar to *upc_memcpy_nb* extension in UPC 1.3





UPC++ Translation Example



Case 1: access local data

- 1. Get the partition id of the global address (1 cycle)
- 2. Check if the partition is local (1 cycle)
- 3. Get the local address of the partition (1 cycle)
- 4. Access data through the local address (1 cycle)

3 CPU cycles for address translation vs. 1 cycle for real work (Bad: 3X overhead)

Case 2: access remote data

- 1. Get the partition id of the global address (1 cycle)
- 2. Check if the partition is local (1 cycle)
- 3. Get the local address of the partition (1 cycle)
- Access data through the network (~10⁴ cycles)

3 CPU cycles for address translation vs. ~10⁴ cycles for real work (Good: 0.3% overhead)




How to Amortize Address Translation Overheads

• Move data in chunks

```
copy(src, dst, count);
non-blocking async_copy is even better
```

Cast pointer-to-shared to pointer-to-local





Completion Events for Non-blocking Put



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```
async_copy_and_signal(global_ptr<T> src,
    global_ptr<T> dst,
    size_t count,
    event *signal_event,
    event *local_completion,
    event *remote_completion);
```

- Three key events for a non-blocking put
 - -Initiator side events :
 - local completion: the src buffer is reusable
 - remote completion: the data has arrived in the dst buffer
 - -Target side event :
 - signal event: the data has arrived in the dst buffer







UPC++ Cheat Sheet for UPC Programmers

UPC		UPC++	
Num. of threads	THREADS	ranks()	
My ID	MYTHREAD	myrank()	
Shared variable shared Type s		shared_var <type> s</type>	
Shared array	shared [bf] Type A[sz]	<pre>shared_array<type> A A.init(sz, bf) global_ptr<type> ptr</type></type></pre>	
Pointer-to-shared	shared Type *ptr		
Dynamic memory allocation	shared void * upc_alloc(nbytes)	<pre>global_ptr<type> allocate<type>(place, count)</type></type></pre>	
Bulk data transfer	upc_memcpy(dst, src, sz)	<pre>copy<type>(src, dst, count)</type></pre>	
Affinity query upc_threadof(ptr)		ptr.where()	
Synchronization	upc_lock_t	shared_lock	
	upc barrier	barrier()	

A "Compiler-Free" Approach for PGAS



- Leverage C++ standards and compilers
 - Implement UPC++ as a C++ template library
 - C++ templates can be used as a mini-language to extend C++ syntax
- Many new features in C++11
 - E.g., type inference, variadic templates, lambda functions, rvalue references
 - C++ 11 is well-supported by major compilers





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Installing UPC++

• Get source from Bitbucket

git clone https://bitbucket.org/upcxx/upcxx.git

• Get the optional multidimensional arrays package

cd upcxx/include

git clone https://bitbucket.org/upcxx/upcxx-arrays.git

Standard autotools build process

./Bootstrap

Create a separate build directory and cd to it configure --with-gasnet=/path/to/\${conduit}-{seq|par}.mak --prefix=/path/to/install CXX=upc++_backend_compiler make; make install

• UPC++ is preinstalled on NERSC Edison (Cray XC30) export MODULEPATH=\$MODULEPATH:/usr/common/usg/degas/modulefiles

module load upc++

Or

. /usr/common/usg/degas/upcxx/default-intel/bin/upcxx_vars.sh

For details about installation instructions, please see
https://bitbucket.org/upcxx/upcxx/wiki/Installing%20UPC++



Compiling UPC++ Programs

 The upc++ compiler wrapper works like the MPI equivalent mpic++. For example,

```
## compile hello.cpp to hello.o
upc++ -c hello.cpp
## compile hello.cpp and link it to a.out
upc++ hello.cpp
## print the command line that upc++ would
execute
upc++ -show
## print the help message
```

```
upc++ -h
```

 You can also get UPC++ makefile definitions and shell environment variables to customize for your app.

https://bitbucket.org/upcxx/upcxx/wiki/Compiling%20UPC++%20Applications





Running UPC++ Programs

- Run it like a MPI (multi-process) program, for example,
 On systems with MPI installed, mpirun
 On a Cray, aprun
- Use the conduit-specific gasnet spawner
- Commonly used GASNet env variables

Increase the size of the global
partition per rank
export GASNET_MAX_SEGSIZE=256MB

Disable process-shared memory nodes
export GASNET_MAX_SUPERNODE=1





Application Examples

```
// shared uint64_t Table[TableSize]; in UPC
         shared_array<uint64_t> Table(TableSize);
        void RandomAccessUpdate()
         {
           uint64_t ran, i;
           ran = starts(NUPDATE / ranks() * myrank());
           for(i = myrank(); i < NUPDATE; i += ranks()) {</pre>
Main
             ran = (ran << 1) ^ ((int64_t)ran < 0 ? POLY : 0);
Table[ran & (TableSize-1)] ^= ran;
update
loop
           }
          Global data layout
                   2
                      3
                              5
                                         8
                                             9
                                                    11
                                                       12
                                                           13 14
           0
               1
                          4
                                  6
                                                10
                                                                  15
          local data layout
                                          Rank 1
                                                             13
              Rank 0
                                 12
                                                      5
                                                         9
                       0
                          4
                              8
              Rank 2
                       2
                          6
                              10
                                 14
                                          Rank 3
                                                  3
                                                     7
                                                         11
                                                            15
                                                                      47
```

Random Access Performance (GUPS)



Performance difference is negligible at large scale





BoxLib

A Software Framework for Block-Structured AMR Applications

Used in many active research projects:

- MAESTRO low Mach number astrophysics
- CASTRO compressible radiation/hydrodynamics
- Nyx cosmology (baryon plus dark matter evolution)
- LMC low Mach number combustion
- CNSReact compressible reacting flow
- ACTuARy atmospheric chemical transport
- PMAMR subsurface modeling (AMANZI-S)

Source: "BoxLib: A Software Framework for Block-Structured AMR Applications" by Ann Almgren http://www.speedup.ch/workshops/w42_2013/ann.pdf







Comm. Patterns in BoxLib

Each process does the following:

```
// Pack data and figure out
// communication neighbors
```

```
MPI_Irecv(...);
MPI_Irecv(...);
...
MPI_Isend(...);
MPI_Isend(...);
```

....

```
// Local computation for
overlap
```

MPI_Waitall(...);

// Unpack data and continue



Cells in each box are stored in column- major order. Boxes are laid out in Z-order in 3D space. Each processor gets a contiguous chunk of boxes of equal size.











Active Receive (Sender Side Message Matching)



- Message matching is done at the sender
- The sender uses signaling put to transfer the message payload and notify the receiver for completion
- The completion event is like a semaphore and can be used to count multiple operations



BoxLib Communication Performance

SMC benchmark on Edison, 128 processes, 1 process per numa node, 12 openmp threads per process

	MPI w. OpenMP	UPC++ w. OpenMP
No	Total Time : 8.2	Total Time : 7.9
overlap	Communication time: 1.8	Communication time: 1.6
	Chemistry time: 2.6	Chemistry time: 2.6
	Hyp-Diff time: 3.8	Hyp-Diff time: 3.8
Overlap	Total Time : 8.4	Total Time : 7.8
	Communication time: 1.8	Communication time: 1.3
	Chemistry time: 2.9	Chemistry time: 2.8
	Hyp-Diff time: 3.8	Hyp-Diff time: 3.8





Progress thread in UPC++

 Mitigate CPU inattentiveness for better communication and computation overlaps

```
progress_thread_start()
progress_thread_stop()
```

- Three threading modes
 - Non thread-safe: main thread explicitly transfers progress control to the progress thread and stop it before making UPC++ calls
 - -Thread-safe with GASNet PAR mode: will need nonthread-specific handle support from GASNet-EX to match the UPC++ usage model
 - -Thread-safe with pthread mutex and GASNet SEQ mode: use a coarse-grained lock for gasnet calls





Application: Full-Waveform Seismic Imaging

- Method for developing models of earth structure, applicable to ...
 - basic science: study of interior structure and composition
 - petroleum exploration and environmental monitoring
 - nuclear test-ban treaty verification
- Model is trained to predict (via numerical simulation) seismograms recorded from real earthquakes or controlled sources
- Training defines a non-linear regression problem, solved iteratively



Application: Full-Waveform Seismic Imaging







Alternative implementation: MPI-3 RMA

- Have to "design for" the MPI implementation
 - NERSC Edison (XC30), so using Cray MPICH 7.0.3 (MPICH 3.0.x)
 - Per-accumulate lock / unlock with exclusive locks
 - Faster than shared (with or without single epoch)
- Would another implementation be faster? (possibly, but hard to say ...)
- In any case, similar code complexity to UPC++
 - Weak scaling vs. UPC++
 - Distributed matrix size fixed (180 GB)
 - Dataset size scaled w/ concurrency
 - 64 updates per MPI or UPC++ task + threads in NUMA domain



https://github.com/swfrench/convergent-matrix-mpi





Scientific results: A whole-mantle model

Geophysical Journal International

Geophys. J. Int. (2014) **199,** 1303–1327 GJI Seismology doi: 10.1093/gji/ggu334

Whole-mantle radially anisotropic shear velocity structure from spectral-element waveform tomography

- First-ever whole-mantle seismic model from numerical waveform tomography
- Reveals new details of deep structure not seen before
- Made feasible by Gauss-Newton scheme, enabled by UPC++



Right: Broad plumes in the earth's lower mantle, including those beneath Pitcairn, Samoa, and other hotspots.

Left: 3D rendering of low-velocity structure beneath the Hawaii hotspot.

(French and Romanowicz, 2015, in revision)



Alternative implementation: MPI-3 RMA

Why the performance disparity?

- Very different approaches to achieving "generality"
- Determines what optimizations are available to programmer

upcxx::async		MPI_Accumulate	
	(general <i>functions</i>)	5.	(general <i>data types</i>)
•	Explicit buffer management	•	Opaque internal MPI buffers
•	Customized update function with domain knowledge	•	Generalized MPI data types + pre-defined merge ops
•	Progress at both source and target is controllable	•	Progress is implspecific and not controllable at target
•	One way bulk data movement can be guaranteed	•	Data may take more than one trip to ensure passive target (ex: bulk accumulate in foMPI)



More opportunities to exploit problem / domain specific knowledge



Multidimensional Arrays in UPC++

Multidimensional Arrays

- Multidimensional arrays are a common data structure in HPC applications
- However, they are poorly supported by the C family of languages, including UPC
 - -Layout, indexing must be done manually by the user
 - -No built-in support for subviews
- Remote copies of array subsets pose an even greater problem
 - -Require manual packing at source, unpacking at destination
 - In PGAS setting, remote copies that are logically one-sided require two-sided coordination by the user





UPC++ Multidimensional Arrays

- True multidimensional arrays with sizes specified at runtime
- Support subviews without copying (e.g. view of interior)
- Can be created over any rectangular index space, with support for strides
- Local-view representation makes locality explicit and allows arbitrarily complex distributions
 - -Each rank creates its own piece of the global data structure
- Allow fine-grained remote access as well as one-sided bulk copies





UPC++ Arrays Based on Titanium

- Titanium is a PGAS language based on Java
- Line count comparison of Titanium and other languages:





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Titanium vs. UPC++

- Main goal: provide similar productivity and performance as Titanium in UPC++
- Titanium is a language with its own compiler
 - -Provides special syntax for indices, arrays
 - PhD theses have been written on compiler optimizations for multidimensional arrays (e.g. Geoff Pike specifically for Titanium)
- Primary challenge for UPC++ is to provide Titanium-like productivity and performance in a library
 - -Use macros, templates, and operator/function overloading for syntax
 - -Provide specializations for performance





Overview of UPC++ Array Library

- A point is an index, consisting of a tuple of integers
 point<2> lb = {{1, 1}}, ub = {{20, 10}};
- A rectangular domain is an index space, specified with a lower bound, upper bound, and optional stride
 rectdomain<2> r(lb, ub); [1,1]

[20,10]

- An array is defined over a rectangular domain and indexed with a point
 ndarray<double, 2> A(r); A[lb] = 3.14;
- One-sided copy operation copies all elements in the intersection of source and destination domains ndarray<double, 2, global> B = ...;
 B.async_copy(A); // copy from A to B async_wait(); // wait for copy completion

Multidimensional Arrays in UPC++ (and Titanium)

• Titanium arrays have a rich set of operations



- None of these modify the original array, they just create another view of the data in that array
- You create arrays with a RectDomain and get it back later using A.domain() for array A
 - A Domain is a set of points in space
 - A RectDomain is a rectangular one
- Operations on Domains include +, -, * (union, different intersection)



Example: 3D 7-Point Stencil



Arrays in Adaptive Mesh Refinement

- AMR starts with a coarse grid over the entire domain
- Progressively finer AMR levels added as needed over subsets of the domain
- Finer level composed of union of regular subgrids, but union itself may not be regular
- Individual subgrids can be represented with UPC++ arrays



Directory structure can be used to represent union of all subgrids



Example: Ghost Exchange in AMR



Syntax of Points

- A point<N> consists of N coordinates
- The point class template is declared as plain-old data (POD), with an N-element array as its only member template<int N> struct point { cint_t x[N];

};

-Can be constructed using initializer list

 $point<2> lb = \{\{1, 1\}\};\$

• The PT function creates a point in non-initializer contexts

point < 2 > 1b = PT(1, 1);

-Implemented using variadic templates in C++11, explicit overloads otherwise



Array Template

 Arrays represented using a class template, with element type and dimensionality arguments

- Last two (optional) arguments specify locality and layout
 - -Locality can be local (i.e. elements are located in the local memory space) or global (elements may be located elsewhere)
 - -Layout can be **strided**, **unstrided**, **row**, **column**; more details later
- Template metaprogramming used to encode type lattices for implicit conversions



Array Implementation

- Local and global arrays have significant differences in their implementation
 - -Global arrays may require communication
- Layout only affects indexing
- Implementation strategy:



 Macros and template metaprogramming used to interface between layers


Foreach Implementation

- Macros and templates allow definition of foreach loops
- C++11 implementation using type inference and lambda:
 #define foreach(p, dom)

```
foreach_(p, dom, UNIQUIFYN(foreach_ptr_, p))
```

 Pre-C++11 implementation also possible using sizeof operator

-However, loop is flattened, so performance is much slower



C++11 Foreach Translation

- Lambda encapsulates body, passed to loop template
 template<int N> struct rditer {
 template<class F> rditer &operator=(const F &func) {
 rdloop<N>::loop(func, p0.x, p1.x, loop_stride.x);
 return *this;
 }
 };
- Loop template implemented recursively, with base case a template specialization that calls body (not shown)



Layout Specializations

- Arrays can be created over any logical domain, but are laid out contiguously
 - -Physical domain may not match logical domain
 - Non-matching stride requires division to get from logical to physical
 - (px[0] base[0])*side_factors[0]/stride[0] +
 - (px[1] base[1])*side_factors[1]/stride[1] +
 - (px[2] base[2])*side_factors[2]/stride[2]
- Introduce template specializations to restrict layout

 strided: any logical or physical stride
 unstrided: logical and physical strides match
 row: matching strides + row-major format
 Default in UPC++ to provide best performance

 (++-column: matching strides + column-major



Array Library Evaluation

- Evaluation of array library done by porting benchmarks from Titanium to UPC++
 - -Again, goal is to match Titanium's productivity and performance without access to a compiler
- Benchmarks: 3D 7-point stencil, NAS CG, FT, and MG
- Minimal porting effort for these examples, providing some evidence that productivity is similar to Titanium
 - -Less than a day for each kernel
 - -Array code only requires change in syntax
 - -Most time spent porting Java features to C++





NAS Benchmarks on One Node (GCC)



Stencil Weak Scaling (GCC)



Array Library Summary

- We have built a multidimensional array library for UPC++
 - -Macros and template metaprogramming provide a lot of power for extending the core language
 - -UPC++ arrays can provide the same productivity gains as Titanium
 - Specializations allow UPC++ to match Titanium's performance
- Some issues remain
 - -Improve performance of one-sided array copies
 - Performance somewhat slower than manual packing/unpacking, as will be shown in miniGMG results
 - -GCC and Clang optimize complex template code well, but other compilers do not



• We are not the only ones to run into this (e.g. Raja, HPX)

Case Study: miniGMG

- We evaluate the productivity and performance of three implementations of **miniGMG**, a multigrid benchmark
- The three implementations use different communication strategies enabled by the PGAS model
 - 1. Fine-grained communication, at the natural granularity of the algorithm
 - 2. Bulk communication, with manual packing and unpacking by the user
 - One-sided analogue of message passing
 - 3. Higher-level array-based communication that offloads the work to an array library
 - Still semantically one-sided

•We evaluate performance on two current platforms



Multigrid Overview

- Linear Solvers (Ax=b) are ubiquitous in scientific computing
 Combustion, Climate, Astrophysics, Cosmology, etc.
- Multigrid exploits the nature of elliptic PDEs to provide a hierarchical approach with O(N) computational complexity
- **Geometric Multigrid** is specialization in which the linear operator (A) is simply a stencil on a structured grid (i.e. *matrix-free*)







miniGMG Overview

- 3D Geometric Multigrid benchmark designed to proxy MG solves in BoxLib and CHOMBO-based AMR applications
- Defines a cubical problem domain
 - Decomposed into cubical subdomains (boxes)
 - Rectahedral collections of subdomains are assigned to processes
 - Decomposition preserved across all levels of V-Cycle
- MPI+OpenMP parallelization
- Configured to use...
 - Fixed 10 U-Cycles (V-Cycle truncated when boxes are coarsened to 4³)
 - 7-pt stencil with Gauss Seidel Red-Black (GSRB) smoother that requires nearest-neighbor communication for each smooth or residual calculation.
 - BiCGStab coarse-grid (bottom) solver
- Communication pattern is thus...
 - Fixed 6 nearest-neighbor communication
 - Message sizes vary greatly as one descends through the V-Cycle (128KB -> 128 bytes -> 128KB)
 - + Requires neighbor synchronization on each step (e.g. two-sided MPI)





Array Creation in miniGMG



Communication Setup for miniGMG Arrays

Compute Intersection

rectdomain<3> isct = dst.domain()*src.domain().shrink(ghosts);

Save Views of Source and Destination Restricted to Intersection
send_arrays[PT(level, g, nn, i, j, k)] = src.constrict(isct);
recv_arrays[PT(level, g, nn, i, j, k)] = dst.constrict(isct);



Bulk Communication Strategy

- Bulk version uses manual packing/unpacking
 - -Similar to MPI code, but with one-sided puts instead of two-sided messaging







Fine-Grained Communication Strategy

- Fine-Grained version does multiple one-sided puts of contiguous data
 - -Puts are at natural granularity of the algorithm







Array Communication Strategy

• Array version logically copies entire ghost zones, delegating actual procedure to array library

-Copies have one-sided semantics







Communication Coordination

- Shared array used to coordinate communication shared array<global ptr<subdomain type>, 1> global_boxes;
- Bulk version must carefully coordinate send and receive buffers between ranks
 - -Must ensure right buffers are used, same ordering for packing and unpacking elements
 - -Special cases for ghost zones at faces, edges, and corners
 - -Most difficult part of code
- Minimal coordination required for fine-grained and array
- -Only need to obtain location of target grid from shared array





Ghost-Zone Exchange Algorithms

	Bulk	Fine-Grained	Array
Barrier	Yes	Yes	Yes
Pack	Yes	Νο	Νο
Async Puts/Copies	1 per neighboring rank	1 for each contiguous segment	1 per neighboring grid
Async Wait	Yes	Yes	Yes
Barrier	Yes	Yes	Yes
Unpack	Yes	Νο	Νο
~ Line Count of Setup + Exchange	884	537	399

Pack/unpack parallelized using OpenMP in bulk version

Up(+ – Effectively serialized in fine-grained and array



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Bulk Copy Code

• Packing/unpacking code in bulk version:

```
for (int k = 0; k < dim_k; k++) {</pre>
  for (int j = 0; j < dim_j; j++) {</pre>
    for (int i = 0; i < dim_i; i++) {</pre>
      int read_ijk = (i+ read_i) + (j+ read_j)*
         read_pencil + (k+ read_k)* read_plane;
      int write_ijk = (i+write_i) + (j+write_j)*
        write_pencil + (k+write_k)*write_plane;
      write[write_ijk] = read[read_ijk];
}
```

Code must be run on both sender and receiver



Fine-Grained Copy Code

 Fine-grained code matches shared-memory code, but with async_copy instead of memcpy:

```
for (int k = 0; k < dim_k; k++)
for (int j = 0; j < dim_j; j++) {
    int roff = recv_i + (j+recv_j)*rpencil +
        (k+recv_k)*rplane;
    int soff = send_i + (j+send_j)*spencil +
        (k+send_k)*splane;
        async_copy(sbuf+soff, rbuf+roff, dim_i);
    }
}</pre>
```

 Takes advantage of fact that source and destination layouts match



- Array version delegates actual copies to array library:
 rcv = recv_arrays[PT(level, g, nn, i, j, k)];
 rcv.async_copy(send_arrays[PT(level, g, nn, i, j, k)]);
- Array library behavior for cases that occur in miniGMG:
 - 1. If the source and destination are contiguous, then one-sided put directly transfers data
 - 2. Otherwise, elements packed into contiguous buffer on source
 - a) If the elements and array metadata fit into a medium active message (AM), a medium AM is initiated
 - AM handler on remote side unpacks into destination
 - b) Otherwise, a short AM is used to allocate a remote buffer
 - Blocking put transfers elements to remote buffer
 - Medium AM transfers array metadata
 - AM handler on remote side unpacks and deallocates buffer



Platforms and Experimental Setup

- Cray XC30 (Edison), located at NERSC
 - -Cray Aries Dragonfly network
 - -Each node has two 12-core sockets
 - -We use 8 threads/socket
- IBM Blue Gene/Q (Mira), located at Argonne
 - -5D torus network
 - -Each node has 16 user cores, with 4 threads/core
 - -We use 64 threads/socket
- Fixed (weak-scaling) problem size of 128³ grid/socket
- Two experiments on each platform
 - -1 MPI process, 8 or 64 OpenMP threads per socket

-8 MPI processes, 1 or 8 OpenMP threads per socket



Communication Histogram

 Histogram of message sizes per process, when using 1 process/socket, for all three versions on Cray XC30



1 Process/Socket, 128³/Process



Histogram of 1 MPI Process vs. 8/Socket

- Same overall problem size per socket
- Fewer small messages per process when using 8 processes, but more small messages per socket



Performance Results on Cray XC30

- Fine-grained and array versions do much better with higher injection concurrency
 - Array version does not currently parallelize packing/unpacking, unlike bulk/MPI



Performance Results on IBM Blue Gene/Q

- Fine-grained does worse, array better on IBM than Cray
- Using more processes improves fine-grained and array performance, but fine-grained still significantly slower



miniGMG Summary

- Array abstraction can provide better productivity than even fine-grained, shared-memory-style code, while getting close to bulk performance
 - Unlike bulk, array code doesn't require two-sided coordination
 - Further optimization (e.g. parallelize packing/unpacking) can reduce the performance gap between array and bulk
 - Existing code can be easily rewritten to take advantage of array copy facility, since changes localized to communication part of code





- Many productive language features can be implemented in C++ without modifying the compiler
 - -Macros and template metaprogramming provide a lot of power for extending the core language
- Many Titanium applications can be ported to UPC++ with little effort
 - -UPC++ can provide the same productivity gains as Titanium
- However, analysis and optimization still an open question
 - -Can we build a lightweight standalone analyzer/optimizer for UPC++?
- -Can we provide automatic specialization at runtime in C++?



- Communicate more often
 - use non-blocking one-sided operations
- Move computation instead of data
 - use async and event-driven execution
- Express algorithms with high-level data structures
 - use Titanium-style multidimensional arrays
- Easy on-ramp
 - interoperate w. existing MPI+OpenMP codes
- We look forward to collaboration!
 - share knowledge and experience beyond tools

UPC++: https://bitbucket.org/upcxx



