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UPC++: A High-Performance Communication Framework for Asynchronous Computation

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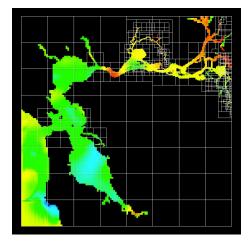
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Some motivating applications

PGAS well-suited to applications that use irregular data structures

- Sparse matrix methods
- Adaptive mesh refinement
- Graph problems, distributed hash tables

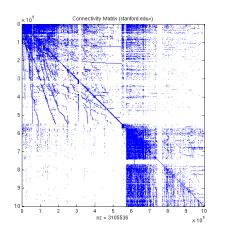


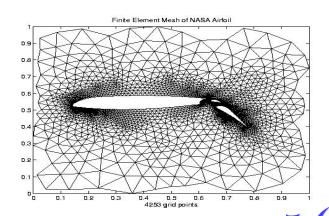
http://tinyurl.com/yxqarenl

Processes may send different amounts of information to other processes

The amount can be data dependent, dynamic

Courtesy of Jim Demmel







The impact of fine-grained communication

The first exascale systems will appear soon

• In the USA: Frontier (2021) https://tinyurl.com/y2ptx3th



Some apps employ fine-grained communication

- Messages are short, so the overhead term dominates communication time $\alpha + F(\beta^{-1}_{\infty}, n)$
- They are latency-limited, and latency is only improving slowly

Memory per core is dropping, an effect that can force more frequent fine-grained communication

We need to reduce communication costs

- Asynchronous communication and execution are critical
- But we also need to keep overhead costs to a minimum



Reducing communication overhead

What if we could let each process directly access one another's memory via a global pointer?

- We don't need to match sends to receives
- We don't need to guarantee message ordering
- There are no unexpected messages

Communication is one-sided

 All metadata provided by the initiator, rather than split between sender and receiver

Looks like shared memory

Observation: modern network hardware provides the capability to directly access memory on another node: Remote Direct Memory Access (RDMA)

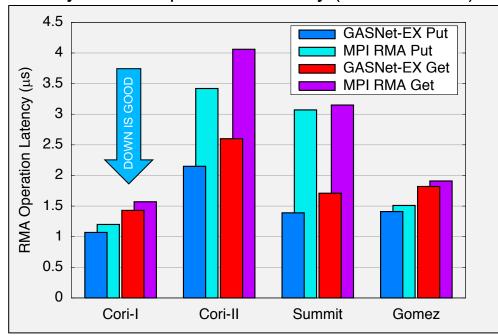
 Can be compiled to load/store if source and target share physical memory



RMA performance: GASNet-EX vs MPI-3

- Three different MPI implementations
- Two distinct network hardware types
- On these four systems the performance of GASNet-EX meets or exceeds MPI RMA:

8-Byte RMA Operation Latency (one-at-a-time)



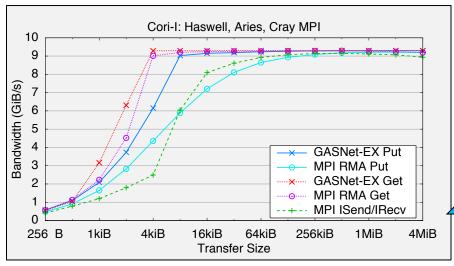
- 8-byte Put latency 6% to 55% better
- 8-byte Get latency 5% to 45% better
- Better flood bandwidth efficiency, typically saturating at ½ or ¼ the transfer size (next slide)

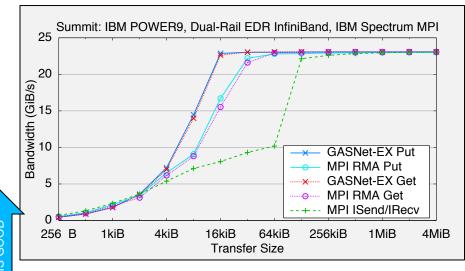
GASNet-EX results from v2018.9.0 and v2019.6.0. MPI results from Intel MPI Benchmarks v2018.1. For more details see Languages and Compilers for Parallel Computing (LCPC'18). https://doi.org/10.25344/S4QP4W More recent results on Summit here replace the paper's results from the older Summitdev.

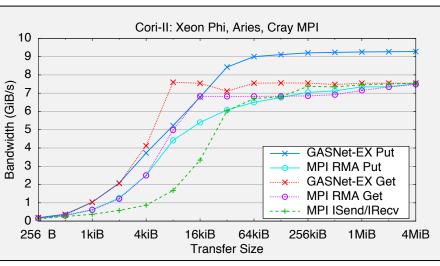


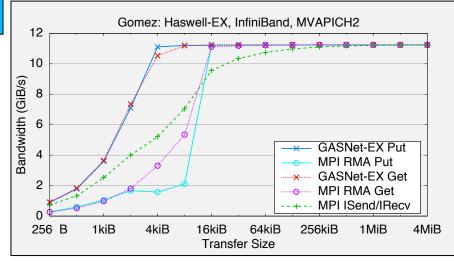
RMA performance: GASNet-EX vs MPI-3

Uni-directional Flood Bandwidth (many-at-a-time)









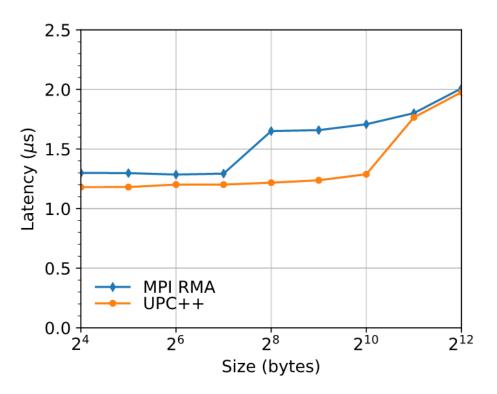




RMA microbenchmarks

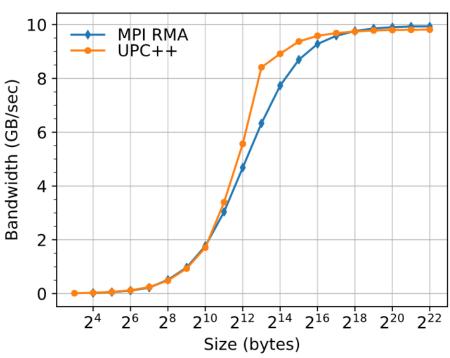
Experiments on NERSC Cori:

Cray XC40 system



Two processor partitions:

- Intel Haswell (2 x 16 cores per node)
- Intel KNL (1 x 68 cores per node)



Round-trip Put Latency (lower is better)

Flood Put Bandwidth (higher is better)

Data collected on Cori Haswell (https://doi.org/10.25344/S4V88H)





The PGAS model

Partitioned Global Address Space

- Support global visibility of storage, leveraging the network's RDMA capability
- Distinguish private and shared memory
- Separate synchronization from data movement

Languages that support PGAS: UPC, Titanium, Chapel, X10, Co-Array Fortran (Fortran 2008)

Libraries that support PGAS: Habanero UPC++, OpenSHMEM, Co-Array C++, Global Arrays, DASH, MPI-RMA

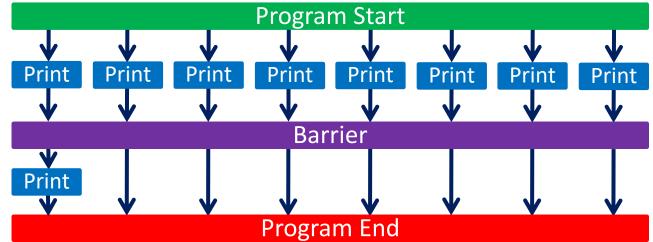
This presentation is about UPC++, a C++ library developed at Lawrence Berkeley National Laboratory



Execution model: SPMD

Like MPI, UPC++ uses a SPMD model of execution, where a fixed number of processes run the same program

```
int main() {
  upcxx::init();
  cout << "Hello from " << upcxx::rank_me() << endl;
  upcxx::barrier();
  if (upcxx::rank_me() == 0) cout << "Done." << endl;
  upcxx::finalize();
}</pre>
```



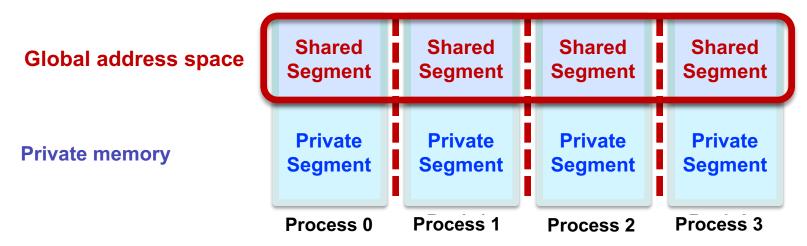
A Partitioned Global Address Space

Global Address Space

- Processes may read and write shared segments of memory
- Global address space = union of all the shared segments

Partitioned

- Global pointers to objects in shared memory have an affinity to a particular process
- Explicitly managed by the programmer to optimize for locality
- In conventional shared memory, pointers do not encode affinity

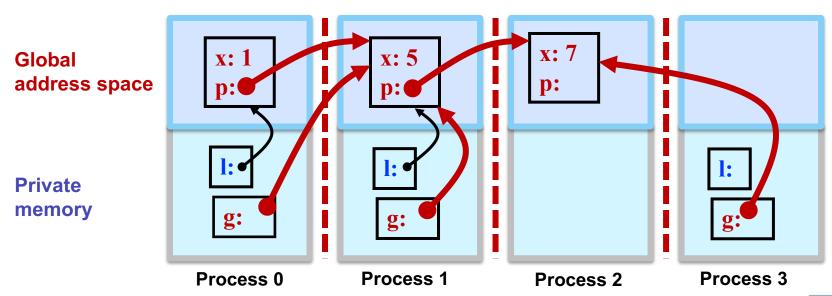




Global vs. raw pointers

We can create data structures with embedded global pointers

Raw C++ pointers can be used on a process to refer to objects in the global address space that have affinity to that process

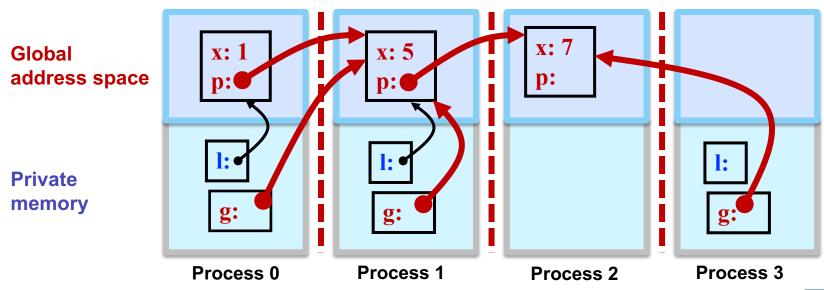


What is a global pointer?

A global pointer carries both an address and the affinity for the data

Parameterized by the type of object it points to, as with a C++ (raw) pointer: e.g. global_ptrdouble>

The affinity identifies the process that created the object





How does UPC++ deliver the PGAS model?

A "Compiler-Free," library approach

 UPC++ leverages C++ standards, needs only a standard C++ compiler





Relies on GASNet-EX for low-overhead communication

- Efficiently utilizes the network, whatever that network may be, including any special-purpose offload support
- Active messages efficiently support Remote Procedure Calls (RPCs), which are expensive to implement in other models
- Enables portability (laptops to supercomputers)

Designed to allow interoperation with existing programming systems

- Same process model as MPI, enabling hybrid applications
- OpenMP and CUDA can be mixed with UPC++ in the same way as MPI+X

What does UPC++ offer?

Asynchronous behavior based on futures/promises

- RMA: Low overhead, zero-copy, one-sided communication.
 Get/put to a remote location in another address space
- RPC: Remote Procedure Call: move computation to the data

Design principles encourage performant program design

- All communication is syntactically explicit
- All communication is asynchronous: futures and promises
- Scalable data structures that avoid unnecessary replication



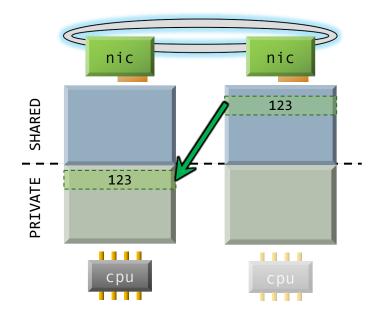
Asynchronous communication

By default, all communication operations are split-phased

- Initiate operation
- Wait for completion
 A future holds a value and a state: ready/not-ready

```
global_ptr<int> gptr1 = ...;
future<int> f1 = rget(gptr1);
// unrelated work...
int t1 = f1.wait();
```

Wait returns the result when the rget completes

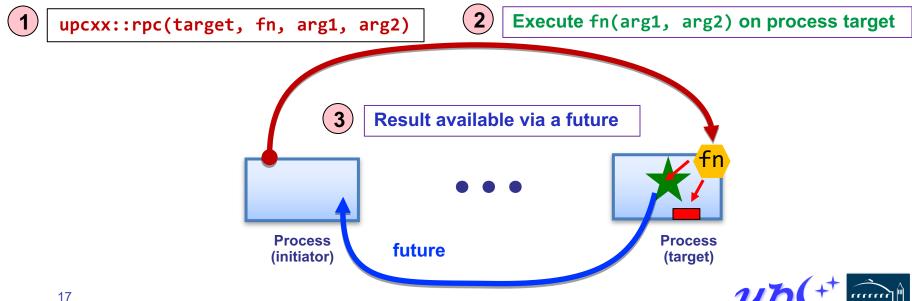




Remote procedure call

Execute a function on another process, sending arguments and returning an optional result

- Initiator injects the RPC to the *target* process
- Target process executes fn(arg1, arg2) at some later time determined at the target
- Result becomes available to the initiator via the future Many RPCs can be active simultaneously, hiding latency



Example: Hello world

```
#include <iostream>
#include <upcxx/upcxx.hpp>
using namespace std;
int main() {
                                              Set up UPC++
  upcxx::init();
                                                runtime
  cout << "Hello world from process</pre>
       << upcxx::rank me()
       << " out of " << upcxx::rank n()
       << " processes" << endl;</pre>
  upcxx::finalize();
                              Close down
                            UPC++ runtime
        Hello world from process 0 out of 4 processes
        Hello world from process 2 out of 4 processes
        Hello world from process 3 out of 4 processes
        Hello world from process 1 out of 4 processes
```



Compiling and running a UPC++ program

UPC++ provides tools for ease-of-use

Compiler wrapper:

- \$ upcxx -g hello-world.cpp -o hello-world.exe
 - Invokes a normal backend C++ compiler with the appropriate arguments (such as -I, -L, -1).
 - We also provide other mechanisms for compiling (upcxx-meta, CMake package).

Launch wrapper:

- \$ upcxx-run -np 4 ./hello-world.exe
 - Arguments similar to other familiar tools
 - We also support launch using platform-specific tools, such as **srun**, **jsrun** and **aprun**.



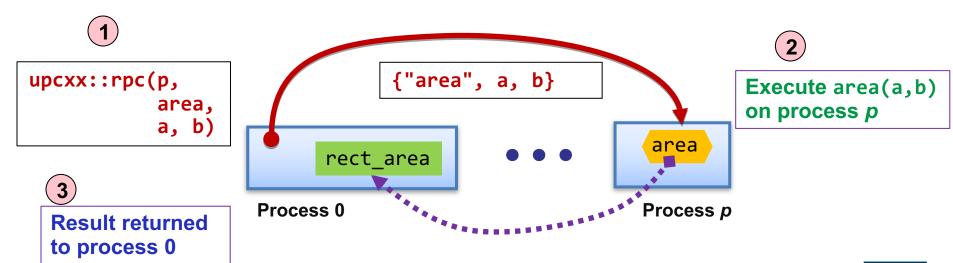
Remote Procedure Calls (RPC)

Let's say that process 0 performs this RPC

```
int area(int a, int b) { return a * b; }
int rect_area = rpc(p, area, a, b).wait();
```

The target process *p* will execute the handler function area() at some later time determined at the target

The result will be returned to process 0





Hello world with RPC (synchronous)

We can rewrite hello world by having each process launch an RPC to process 0

```
int main() {
  upcxx::init();
  for (int i = 0; i < upcxx::rank_n(); ++i) {</pre>
    if (upcxx::rank me() == i) {
                                        C++ lambda function
      upcxx::rpc(0, [](int rank) {
         cout << "Hello from process " << rank << endl;</pre>
       }, upcxx::rank_me()).wait();
                                        Wait for RPC to complete
                                           before continuing
    upcxx::barrier();
                                 Rank number is the
                               argument to the lambda
  upcxx::finalize();
                           Barrier prevents any process from
                           proceeding until all have reached it
```

Futures

RPC returns a *future* object, which represents a computation that may or may not be complete

Calling wait () on a future causes the current process to wait until the future is ready



What is a future?

A future is a handle to an asynchronous operation, which holds:

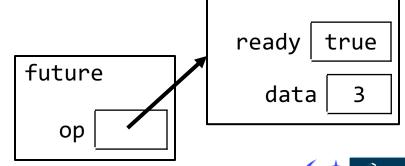
- The status of the operation
- The results (zero or more values) of the completed operation

The future is not the result itself, but a proxy for it

The <u>wait</u>() method blocks until a future is ready and returns the result

```
upcxx::future<int> fut = /* ... */;
int result = fut.wait();
```

The then() method can be used instead to attach a callback to the future





"async op"

Overlapping communication

Rather than waiting on each RPC to complete, we can launch every RPC and then wait for each to complete

```
vector<upcxx::future<int>> results;
for (int i = 0; i < upcxx::rank_n(); ++i) {
   upcxx::future<int> fut = upcxx::rpc(i, []() {
      return upcxx::rank_me();
   }));
   results.push_back(fut);
}

for (auto fut : results) {
   cout << fut.wait() << endl;
}</pre>
```

We'll see better ways to wait on groups of asynchronous operations later



1D 3-point Jacobi in UPC++

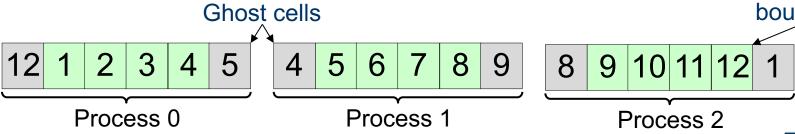
Iterative algorithm that updates each grid cell as a function of its old value and those of its immediate neighbors

Out-of-place computation requires two grids

Local grid size

Sample data distribution of each grid (12 domain elements, 3 ranks, N=12/3+2=6):

Periodic boundary



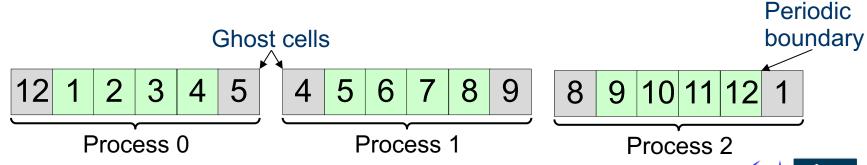


Jacobi boundary exchange (version 1)

RPCs can refer to static variables, so we use them to keep track of the grids

```
double *old_grid, *new_grid;
double get_cell(long i) {
   return old_grid[i];
}
...
double val = rpc(right, get_cell, 1).wait();
```

* We will generally elide the upcxx:: qualifier from here on out.



Jacobi computation (version 1)

We can use RPC to communicate boundary cells

```
future<double> left_ghost = rpc(left, get_cell, N-2);
future<double> right ghost = rpc(right, get cell, 1);
for (long i = 2; i < N - 2; ++i)
  new grid[i] = 0.25 *
    (old_grid[i-1] + 2*old_grid[i] + old_grid[i+1]);
new grid[1] = 0.25 *
  (left_ghost.wait() + 2*old_grid[1] + old_grid[2]);
new grid[N-2] = 0.25 *
  (old_grid[N-3] + 2*old_grid[N-2] + right_ghost.wait());
std::swap(old grid, new grid);
                                           6
                                                 8
                                                    9
```



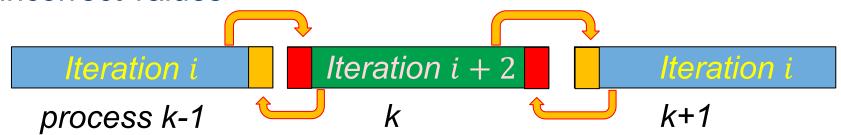
Process 1

Race conditions

Since processes are unsynchronized, it is possible that a process can move on to later iterations while its neighbors are still on previous ones

 One-sided communication decouples data movement from synchronization for better performance

A *straggler* in iteration i could obtain data from a neighbor that is computing iteration i + 2, resulting in incorrect values



This behavior is unpredictable and may not be observed in testing



Naïve solution: barriers

Barriers at the end of each iteration provide sufficient synchronization

```
future<double> left_ghost = rpc(left, get_cell, N-2);
future<double> right ghost = rpc(right, get cell, 1);
for (long i = 2; i < N - 2; ++i)
 /* · · · */:
new grid[1] = 0.25 *
  (left_ghost.wait() + 2*old_grid[1] + old_grid[2]);
new grid[N-2] = 0.25 *
  (old grid[N-3] + 2*old grid[N-2] + right ghost.wait());
barrier();
                                      Barriers around the swap
std::swap(old_grid, new_grid);
                                    ensure that incoming RPCs in
barrier();
                                    both this iteration and the next
                                      one use the correct grids
```



One-sided put and get (RMA)

UPC++ provides APIs for one-sided puts and gets

Implemented using network RDMA if available – most efficient way to move large payloads

Scalar put and get:

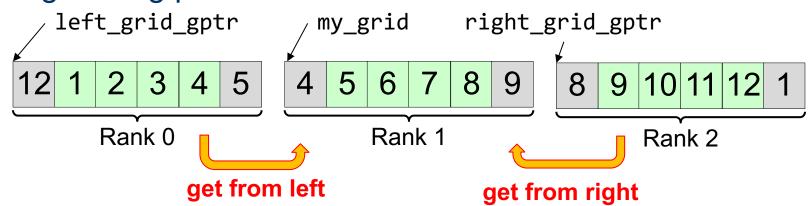
```
global_ptr<int> remote = /* ... */;
future<int> fut1 = rget(remote);
int result = fut1.wait();
future<> fut2 = rput(42, remote);
fut2.wait();
```

Vector put and get:

```
int *local = /* ... */;
future<> fut3 = rget(remote, local, count);
fut3.wait();
future<> fut4 = rput(local, remote, count);
fut4.wait();
```

Jacobi with ghost cells

Each process maintains *ghost cells* for data from neighboring processes



Assuming we have *global pointers* to our neighbor grids, we can do a one-sided put or get to communicate the ghost data:

```
double *my_grid;
global_ptr<double> left_grid_gptr, right_grid_gptr;
my_grid[0] = rget(left_grid_gptr + N - 2).wait();
my_grid[N-1] = rget(right_grid_gptr + 1).wait();
```



Storage management

Memory must be allocated in the shared segment in order to be accessible through RMA

```
global_ptr<double> old_grid_gptr, new_grid_gptr;
...
old_grid_gptr = new_array<double>(N);
new grid gptr = new_array<double>(N);
```

These are <u>not</u> collective calls - each process allocates its own memory, and there is no synchronization

 Explicit synchronization may be required before retrieving another process's pointers with an RPC

UPC++ does not maintain a symmetric heap

 The pointers must be communicated to other processes before they can access the data



Downcasting global pointers

If a process has direct load/store access to the memory referenced by a global pointer, it can *downcast* the global pointer into a raw pointer with local()

```
global_ptr<double> old_grid_gptr, new_grid_gptr;
double *old_grid, *new_grid;

void make_grids(size_t N) {
  old_grid_gptr = new_array<double>(N);
  new_grid_gptr = new_array<double>(N);
  old_grid = old_grid_gptr.local();
  new_grid = new_grid_gptr.local();
}
```

Later, we will see how downcasting can be used with processes that share physical memory



Can be accessed

by an RPC

Jacobi RMA with gets

Each process obtains boundary data from its neighbors with rget()

```
future<>> left get = rget(left old grid + N - 2,
                         old grid, 1);
future<> right get = rget(right old grid + 1,
                          old grid + N - 1, 1);
for (long i = 2; i < N - 2; ++i)
 /* ... */:
left get.wait();
new grid[1] = 0.25 *
  (old grid[0] + 2*old grid[1] + old grid[2]);
right get.wait();
new grid[N-2] = 0.25 *
  (old grid[N-3] + 2*old grid[N-2] + old grid[N-1]);
```



Callbacks

The <u>then</u>() method attaches a callback to a future

 The callback will be invoked after the future is ready, with the future's values as its arguments

```
future<> left update =
  rget(left old grid + N - 2, old grid, 1)
  .<u>then</u>([]() { ← ____
                               Vector get does not produce a value
    new grid[1] = 0.25 *
      (old grid[0] + 2*old grid[1] + old grid[2]);
  });
future<> right update =
  rget(right old grid + N - 2)
  .<u>then</u>([](double value) { ←
                                        Scalar get produces a value
    new grid[N-2] = 0.25 *
      (old grid[N-3] + 2*old grid[N-2] + value);
  });
```



Chaining callbacks

Callbacks can be chained through calls to then()

```
global ptr<int> source = /* ... */;
global ptr<double> target = /* ... */;
future<int> fut1 = rget(source);
future<double> fut2 = fut1.then([](int value) {
    return std::log(value);
});
future<> fut3 =
    fut2.then([target](double value) {
        return rput(value, target);
    });
fut3.wait();
```

This code retrieves an integer from a remote location, computes its log, and then sends it to a different remote location



Conjoining futures

Multiple futures can be *conjoined* with when_all() into a single future that encompasses all their results

Can be used to specify multiple dependencies for a callback

```
global ptr<int> source1 = /* ... */;
global ptr<double> source2 = /* ... */;
global ptr<double> target = /* ... */;
future<int> fut1 = rget(source1);
future < double > fut2 = rget(source2);
future<int, double> both =
                                              rget
                                                        rget
  when all(fut1, fut2);
future<> fut3 =
                                                 when all
  both.then([target](int a, double b) {
    return rput(a * b, target);
  });
                                      then({rput(a*b,target)})
fut3.wait();
```



Jacobi RMA with puts and conjoining

Each process sends boundary data to its neighbors with rput(), and the resulting futures are conjoined

```
future<> puts = when all(
    rput(old grid[1], left old grid + N - 1),
    rput(old grid[N-2], right old grid));
for (long i = 2; i < N - 2; ++i)
 /* · · · */:
                   Ensure outgoing puts have completed
puts.wait();
                   Ensure incoming puts have completed
barrier();
new_grid[1] = 0.25 *
  (old grid[0] + 2*old grid[1] + old grid[2]);
new grid[N-2] = 0.25 *
  (old grid[N-3] + 2*old grid[N-2] + old grid[N-1]);
```



Distributed objects

A distributed object is an object that is partitioned over a set of processes

```
dist_object<T>(T value, team &team = world());
```

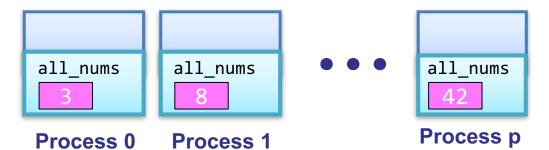
The processes share a universal name for the object, but each has its own local value

Similar in concept to a co-array, but with advantages

- Scalable metadata representation
- Does not require a symmetric heap
- No communication to set up or tear down
- Can be constructed over teams

39

```
dist_object<int>
  all_nums(rand());
```



Bootstrapping the communication

Since allocation is not collective, we must arrange for each process to obtain pointers to its neighbors' grids

We can use a distributed object to do so

<u>barrier</u>(); ←

Ensures distributed objects are not destructed until all ranks have completed their fetches



Implicit synchronization

The future returned by <u>fetch</u>() is not readied until the distributed object has been constructed on the target, allowing its value to be read

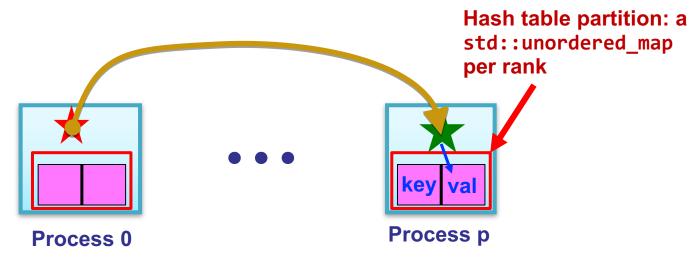
 This allows us to avoid explicit synchronization between the creation and the <u>fetch</u>()



Distributed hash table (DHT)

Distributed analog of std::unordered_map

- Supports insertion and lookup
- We will assume the key and value types are string
- Represented as a collection of individual unordered maps across processes
- We use RPC to move hash-table operations to the owner





DHT data representation

A distributed object represents the directory of unordered maps

```
class DistrMap {
  using dobj_map_t =
    dist object<unordered map<string, string>>;
 // Construct empty map
  dobj_map_t local_map{{}};
                              Computes owner for the given key
  int get_target_rank(const string &key) {
    return std::hash<string>{}(key) % rank n();
```



DHT insertion

Insertion initiates an RPC to the owner and returns a future that represents completion of the insert

```
future<> insert(const string &key,
                      const string &val) {
     return rpc(get_target_rank(key),
        [](dobj_map_t &lmap, const string &key,
           const string &val) {
                                              Key and value passed
          (*lmap)[key] = val;
                                              as arguments to the
        }, local_map, key, val); 
                                                remote function
 UPC++ uses the
distributed object's
universal name to
 look it up on the
 remote process
                                              Process p
                    Process 0
```

DHT find

Find also uses RPC and returns a future

```
future<string> find(const string &key) {
  return rpc(get target rank(key),
    [](dobj_map_t &lmap, const string &key) {
      if (lmap->count(key) == ∅)
        return string("NOT FOUND");
      else
        return (*lmap)[key];
    }, local_map, key);
```



Process p

Process 0

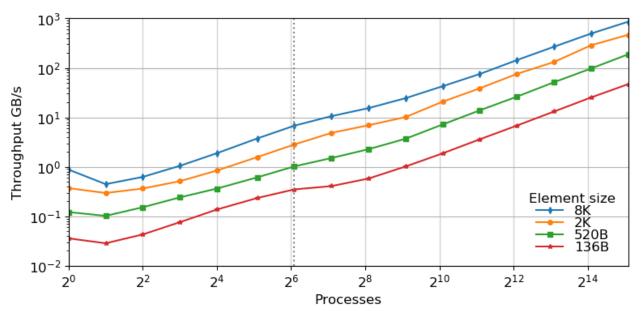
Optimized DHT scales well

Excellent weak scaling up to 32K cores [IPDPS19]

Randomly distributed keys

RPC and RMA lead to simplified and more efficient design

- Key insertion and storage allocation handled at target
- Without RPC, complex updates would require explicit synchronization and two-sided coordination



Cori @ NERSC (KNL) Cray XC40

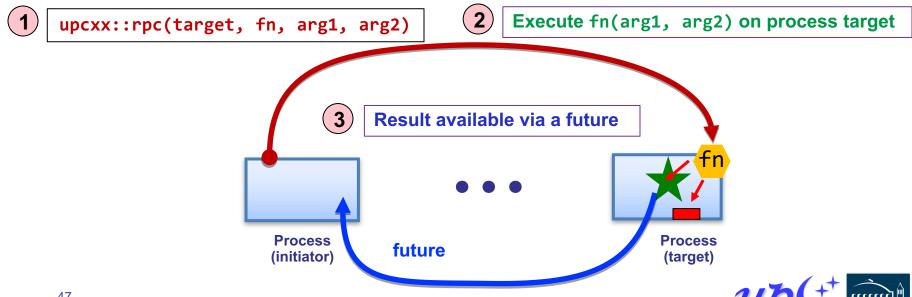


RPC and progress

Review: high-level overview of an RPC's execution

- 1. Initiator injects the RPC to the target process
- 2. Target process executes fn(arg1, arg2) at some later time determined at the target
- 3. Result becomes available to the initiator via the future

Progress is what ensures that the RPC is eventually executed at the target



Progress

UPC++ does not spawn hidden threads to advance its internal state or track asynchronous communication

This design decision keeps the runtime lightweight and simplifies synchronization

 RPCs are run in series on the main thread at the target process, avoiding the need for explicit synchronization

The runtime relies on the application to invoke a progress function to process incoming RPCs and invoke callbacks

Two levels of progress

- Internal: advances UPC++ internal state but no notification
- User: also notifies the application
 - Readying futures, running callbacks, invoking inbound RPCs



Invoking user-level progress

The progress() function invokes user-level progress

So do blocking calls such as wait() and barrier()

A program invokes user-level progress when it expects local callbacks and remotely invoked RPCs to execute

 Enables the user to decide how much time to devote to progress, and how much to devote to computation

User-level progress executes some number of outstanding received RPC functions

- "Some number" could be zero, so may need to periodically invoke when expecting callbacks
- Callbacks may not wait on communication, but may chain new callbacks on completion of communication



Remote atomics

Remote atomic operations are supported with an atomic domain

Atomic domains enhance performance by utilizing hardware offload capabilities of modern networks

The domain dictates the data type and operation set

Support int64_t, int32_t, uint64_t, uint32_t, float, double

Operations are performed on global pointers and are asynchronous

```
global_ptr <int64_t> ptr = new_<int64_t>(0);
future<int64_t> f = dom.fetch_add(ptr,2,memory_order_relaxed);
int64_t res = f.wait();
```



Serialization

RPC's transparently serialize shipped data

- Conversion between in-memory and byte-stream representations
- serialize → transfer → deserialize → invoke target

Conversion makes byte copies for C-compatible types

char, int, double, struct{double;double;}, ...

Serialization works with most STL container types

- vector<int>, string, vector<list<pair<int,float>>>,...
- Hidden cost: containers deserialized at target (copied) before being passed to RPC function



Views

UPC++ *views* permit optimized handling of collections in RPCs, without making unnecessary copies

view<T>: non-owning sequence of elements

When deserialized by an RPC, the <u>view</u> elements can be accessed directly from the internal network buffer, rather than constructing a container at the target



Shared memory hierarchy and local_team

Memory systems on supercomputers are hierarchical

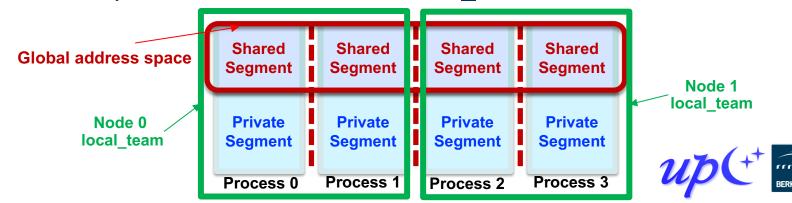
- Some process pairs are "closer" than others
- Ex: cabinet > switch > node > NUMA domain > socket > core

Traditional PGAS model is a "flat" two-level hierarchy

"same process" vs "everything else"

UPC++ adds an intermediate hierarchy level

- local_team() a team corresponding to a physical node
- These processes share a physical memory domain
 - Shared segments are CPU load/store accessible across processes in the same local_team



Downcasting and shared-memory bypass

Earlier we covered downcasting global pointers

Process 1

0 | l id

Process 0

Node 0

local team

- Converting global_ptr<T> from this process to raw C++ T*
- Also works for <u>global_ptr</u><T> from <u>any</u> process in <u>local_team()</u>

```
int l_id = local team().rank me();
                                        Rank and count in my local node
    int l_cnt = local_team().rank_n();
    global ptr<int> gp data;
    if (l_id == 0) gp_data = new array<int>(l_cnt);
                                                       Allocate and share
                                                       one array per node
    gp_data = broadcast(gp_data, 0, local_team()).wait();
    int *lp_data = gp_data.local(); \rightarrow Downcast to get raw C++ ptr to shared array
    Global
                                                                  Node 1
space
                                                                 local team
                           lp data
               lp data
                                          lp data
                                                      lp data
```

0

Process 2

Process 3

54

Optimizing for shared memory in many-core

local_team() allows optimizing co-located processes
for physically shared memory in two major ways:

- Memory scalability
 - Need only one copy per node for replicated data
 - E.g. Cori KNL has 272 hardware threads/node
- Load/store bypass avoid explicit communication overhead for RMA on local shared memory
 - Downcast <u>global_ptr</u> to raw C++ pointer
 - Avoid extra data copies and communication overheads



Completion: synchronizing communication

Earlier we synchronized communication using futures:

```
future<int> fut = rget(remote_gptr);
int result = fut.wait();
```

This is just the default form of synchronization

- Most communication ops take a defaulted completion argument
- More explicit: <u>rget(gptr, operation cx::as_future()</u>);
 - Requests future-based notification of operation completion

Other completion arguments may be passed to modify behavior

- Can trigger different actions upon completion, e.g.:
 - Signal a promise, inject an RPC, etc.
- Can even combine several completions for the same operation

Can also detect other "intermediate" completion steps

For example, source completion of an RMA put or RPC



Completion: promises

A *promise* represents the producer side of an asynchronous operation

A future is the consumer side of the operation

By default, communication operations create an implicit promise and return an associated future

Instead, we can create our own promise and register it with multiple communication operations

```
void do_gets(global_ptr<int> *gps, int *dst, int cnt) {
   promise<> p;
   for (int i = 0; i < cnt; ++i)
        rget(gps[i], dst+i, 1, operation cx::as promise(p));
   future<> fut = p.finalize();
   fut.wait();
}
Close registration
   and obtain an
   associated future
Register an operation
   on a promise
```

Completion: "signaling put"

One particularly interesting case of completion:

- Performs an RMA put, informs the target upon arrival
 - RPC callback to inform the target and/or process the data
 - Implementation can transfer both the RMA and RPC with a single network-level operation in many cases
 - Couples data transfer w/sync like message-passing
 - BUT can deliver payload using RDMA without rendezvous (because initiator specified destination address)



Memory Kinds

Supercomputers are becoming increasingly heterogeneous in compute, memory, storage

UPC++ memory kinds enable sending data between different kinds of memory/storage media

API is meant to be flexible, but initially supports memory copies between remote or local CUDA GPU devices and remote or local host memory

```
global_ptr<int, memory kind::cuda_device> src = ...;
global_ptr<int, memory kind::cuda_device> dst = ...;
copy(src, dst, N).wait();
```

Can point to memory on a local or remote GPU



Non-contiguous RMA

We've seen contiguous RMA

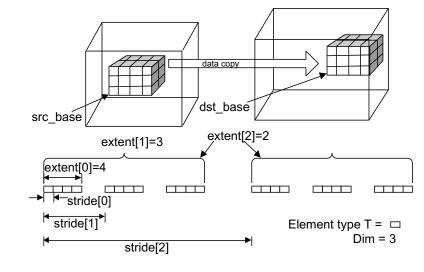
- Single-element
- Dense 1-d array

Some apps need sparse RMA access

- Could do this with loops and fine-grained access
- More efficient to pack data and aggregate communication
- We can automate and streamline the pack/unpack

Three different APIs to balance metadata size vs. generality

- Irregular: *iovec*-style iterators over pointer+length
- Regular: iterators over pointers with a fixed length
- Strided: N-d dense array copies + transposes





UPC++ additional resources

Website: <u>upcxx.lbl.gov</u> includes the following content:

- Open-source/free library implementation
 - Portable from laptops to supercomputers
- Tutorial resources at <u>upcxx.lbl.gov/training</u>
 - UPC++ Programmer's Guide
 - Videos and exercises from past tutorials
- Formal UPC++ specification
 - All the semantic details about all the features
- Links to various UPC++ publications
- Links to optional extensions and partner projects
- Contact information for support

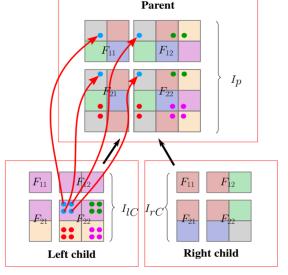


Application case studies

UPC++ has been used successfully in several applications to improve programmer productivity and runtime performance

We discuss two specific applications:

- symPack, a solver for sparse symmetric matrices
- MetaHipMer, a genome assembler







Sparse multifrontal direct linear solver

Sparse matrix factorizations have low computational intensity and irregular communication patterns

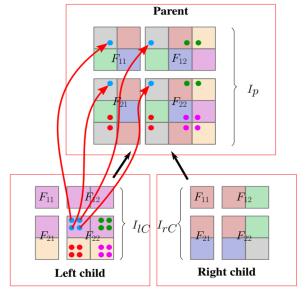
Extend-add operation is an important building block for multifrontal sparse solvers

Sparse factors are organized as a hierarchy of condensed matrices called **frontal matrices**

Four sub-matrices:

factors + contribution block

Code available as part of upcxx-extras BitBucket git repo



Details in IPDPS'19 paper:

Bachan, Baden, Hofmeyr, Jacquelin, Kamil, Bonachea, Hargrove, Ahmed. "UPC++: A High-Performance Communication Framework for Asynchronous Computation", https://doi.org/10.25344/S4V88H



Implementation of the extend-add operation

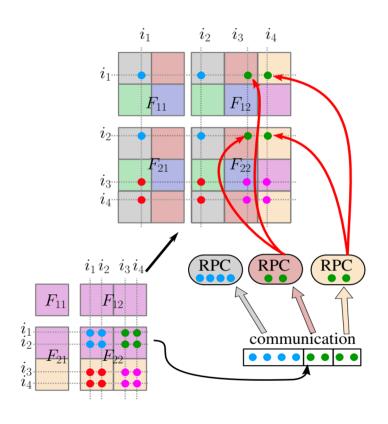
Data is binned into per-destination contiguous buffers

Traditional MPI implementation uses MPI_Alltoallv

 Variants: MPI_Isend/MPI_Irecv + MPI_Waitall/MPI_Waitany

UPC++ Implementation:

- RPC sends child contributions to the parent using a UPC++ view
- RPC callback compares indices and accumulates contributions on the target

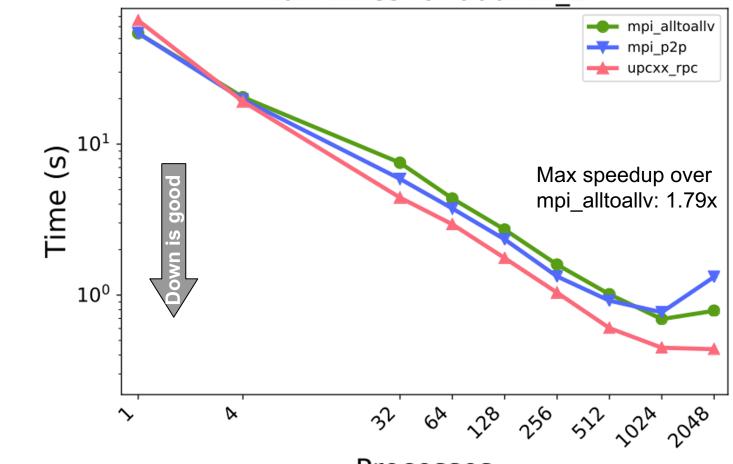




UPC++ improves sparse solver performance

Experiments done on Cori Haswell





Processes

Assembly trees / frontal matrices extracted from STRUMPACK

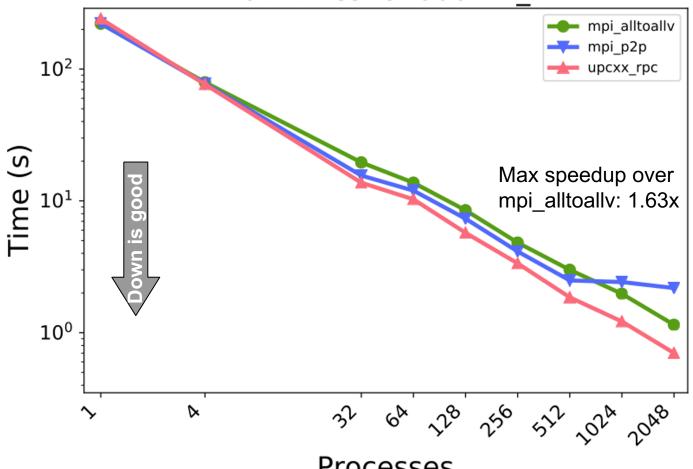
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UPC++ improves sparse solver performance

Experiments done on Cori KNL





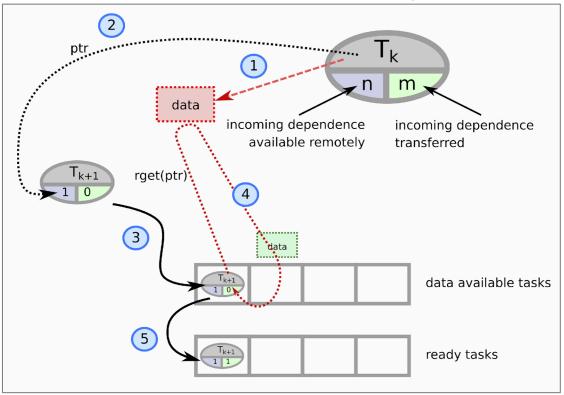
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symPACK: a solver for sparse symmetric matrices



- 1) Data is produced
- 2) Notifications using upcxx::rpc_ff
 - Enqueues a **upcxx::global_ptr** to the data
 - Manages dependency count
- 3) When all data is available, task is moved in the data available task list
- 4) Data is moved using upcxx::rget
 - Once transfer is complete, update dependency count
- When everything has been transferred, task is moved to the ready tasks list



symPACK: a solver for sparse symmetric matrices

Matrix is distributed by supernodes

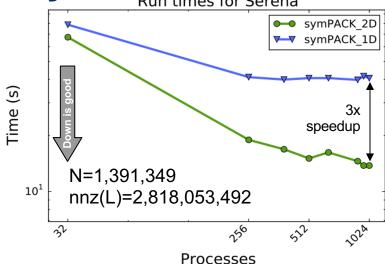
- 1D distribution
 - Balances flops, memory
 - Lacks strong scalability
- New 2D distribution (to appear)
 - Explicit load balancing, not regular block cyclic mapping
 - Balances flops, memory
 - Finer granularity task graph

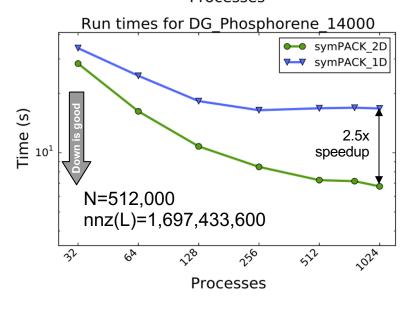
Strong scalability on Cori Haswell:

- Up to 3x speedup for Serena
- Up to 2.5x speedup for DG_Phosphorene_14000

UPC++ enables the finer granularity task graph to be fully exploited

Better strong scalability

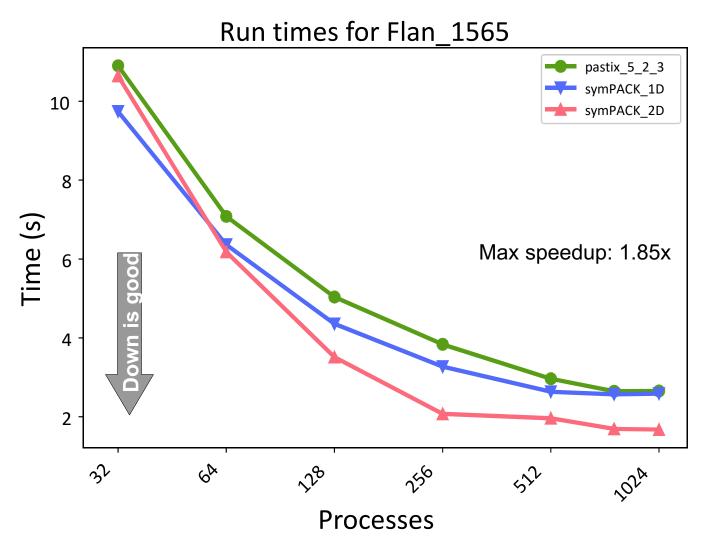








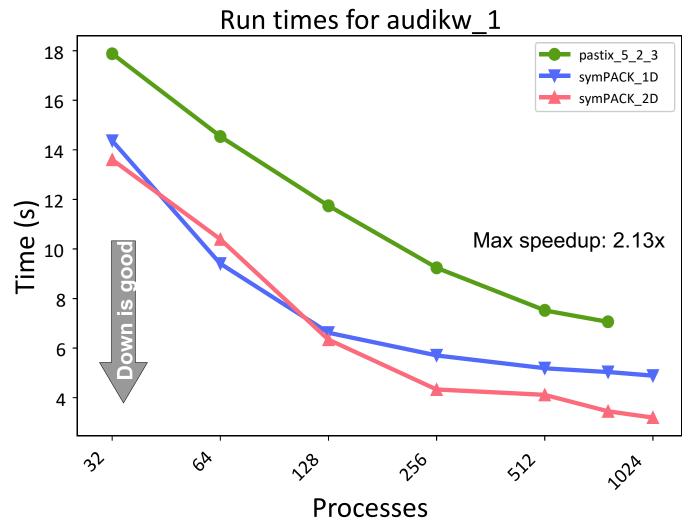
symPACK strong scaling experiment NERSC Cori Haswell



N=1,564,794 nnz(L)=1,574,541,576



symPACK strong scaling experiment NERSC Cori Haswell



N=943,695 nnz(L)=1,261,342,196



UPC++ provides productivity + performance for sparse solvers

Productivity

- RPC allowed very simple notify-get system
- Interoperates with MPI
- Non-blocking API

Reduced communication costs

- Low overhead reduces the cost of fine-grained communication
- Overlap communication via asynchrony and futures
- Increased efficiency in the extend-add operation
- Outperform state-of-the-art sparse symmetric solvers

http://upcxx.lbl.gov

http://sympack.org



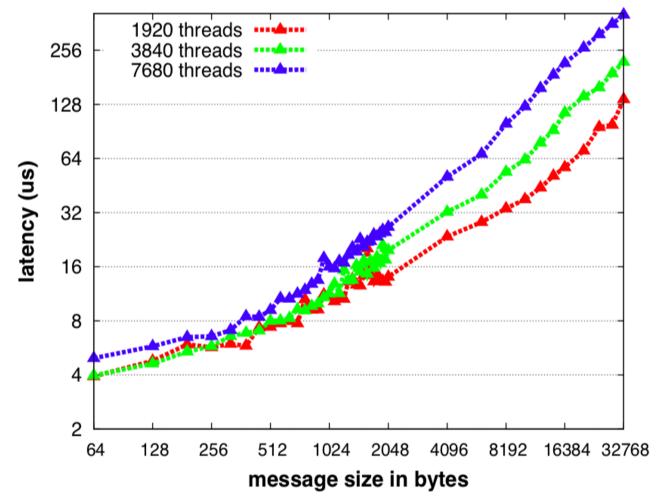
ExaBiome / MetaHipMer distributed hashmap

Memory-limited graph stages

k-mers, contig, scaffolding

Optimized graph construction

 Larger messages for better network bandwidth





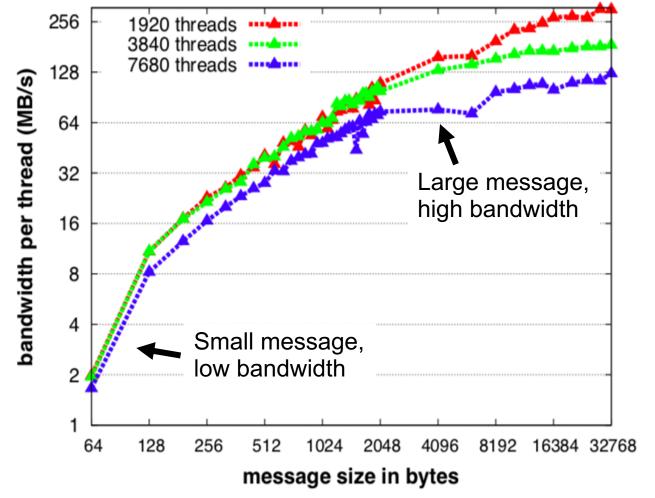
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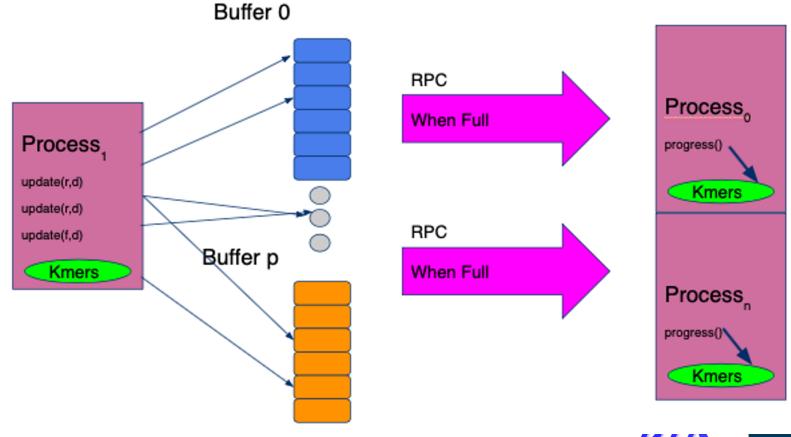




ExaBiome / MetaHipMer distributed hashmap

Aggregated store

- Buffer calls to dist_hash::update(key,value)
- Send fewer but larger messages to target rank





API - AggrStore < FuncDistObject, T>

```
struct FunctionObject {
  void operator()(T &elem) { /* do something */ }
};
using FuncDistObject = upcxx::dist object<FunctionObject>;
// AggrStore holds a reference to func
AggrStore(FuncDistObj &func);
~AggrStore() { clear(); }
// clear all internal memory
void clear();
// allocate all internal memory for buffering
void set_size(size_t max_bytes);
// add one element to the AggrStore
void update(intrank t target rank, T &elem);
// flush and quiesse
void flush updates();
```



MetaHipMer utilized UPC++ features

C++ templates - efficient code reuse

dist_object - as a templated functor & data store

Asynchronous all-to-all exchange - not batch sync

 5x improvement at scale over previous MPI implementation

Future-chained workflow

- Multi-level RPC messages
- Send by node, then by process

Promise & fulfill - for a fixed-size memory footprint

Issue promise when full, fulfill when available



UPC++ additional resources

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