



**Hewlett Packard**  
Enterprise

# **Parallel Programmability from Laptops to HPCs with Chapel and Arkouda**

Brad Chamberlain

UW CSE PLSE

January 28, 2025

**A Bit About Me**





# What is Chapel?

**Chapel:** A modern parallel programming language

- Portable & scalable
- Open-source & collaborative

## Goals:

- Support general parallel programming
- Make parallel programming at scale far more productive



# Productive Parallel Programming: One Definition

Imagine a programming language for parallel computing that is as...  
...**readable and writeable** as Python

...yet also as...

...**fast** as Fortran / C / C++

...**scalable** as MPI / SHMEM

...**GPU-ready** as CUDA / HIP / OpenMP / Kokkos / OpenCL / OpenACC / ...

...**portable** as C

...**fun** as [your favorite programming language]

**This is our motivation for Chapel**





# Six Key Characteristics of Chapel

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1. **portable:** runs on laptops, clusters, the cloud, supercomputers
2. **open-source:** to reduce barriers to adoption and leverage community contributions
3. **compiled:** to generate the best performance possible
4. **statically typed:** to avoid simple errors after hours of execution
5. **interoperable:** with C, C++, Fortran, Python, ...
6. **from scratch:** not a dialect or extension of another language  
(though inspiration was taken from many)



# Outline

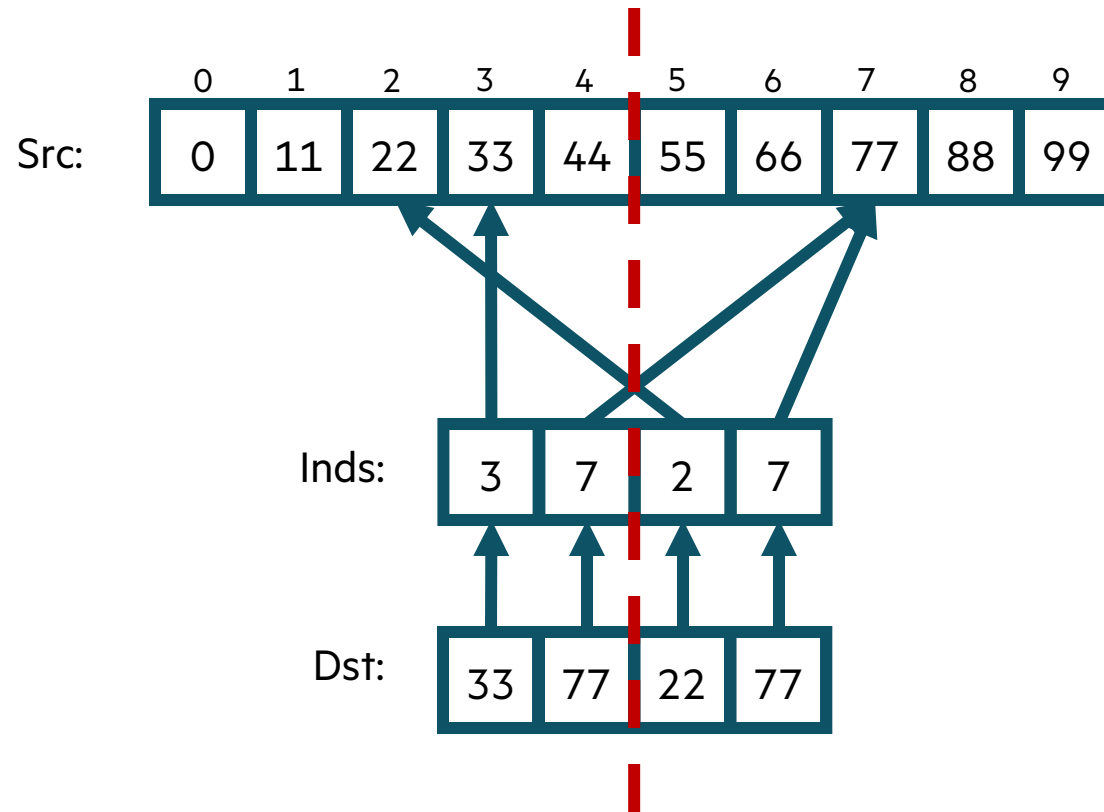
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- **Chapel Goals and Characteristics**
- **A Brief Introduction to Chapel**
- **Applications of Chapel**
- **Global-view vs. SPMD Programming**
- **Chapel Parallelism and Locality Features**
- **Sample Compiler Optimizations**
- **Programming GPUs in Chapel**
- **Wrap-up**

# **A Brief Introduction to Chapel (via Bale IndexGather)**

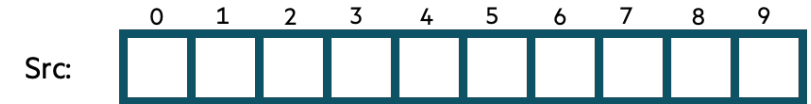


## Bale IndexGather (IG): In Pictures



# Bale IG in Chapel: Array Declarations

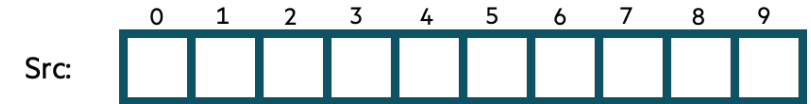
```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    Inds, Dst: [0..  
        int;  
        int;
```



\$

# Bale IG in Chapel: Compiling

```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    Inds, Dst: [0..  
    int;
```



```
$ chpl bale-ig.chpl
```

```
$
```

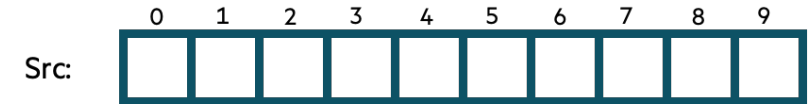




# Bale IG in Chapel: Executing


```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    Inds, Dst: [0..  
    int;
```


```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```




# Bale IG in Chapel: Executing, Overriding Configs

```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    Inds, Dst: [0..  
        int;
```

Src: 

Inds: 

Dst: 

```
$ chpl bale-ig.chpl  
$ ./bale-ig --n=1_000_000 --m=1_000_000  
$
```



# Bale IG in Chapel: Array Initialization

```
use Random;

config const n = 10,
             m = 4;

var Src: [0..
```

```
$ chpl bale-ig.chpl
$ ./bale-ig
$
```

	0	1	2	3	4	5	6	7	8	9
Src:	0	11	22	33	44	55	66	77	88	99

Inds:	3	7	2	7
-------	---	---	---	---

Dst:				
------	--	--	--	--

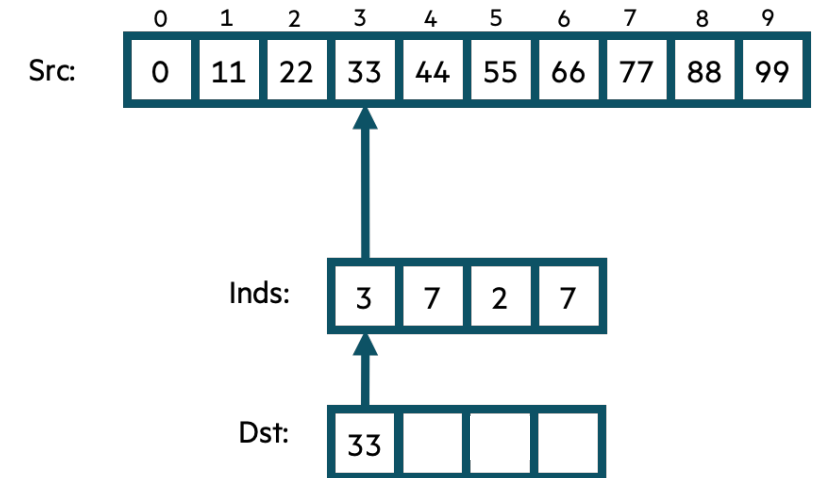




# Bale IG in Chapel: Serial Version

```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    Inds, Dst: [0..  
...  
for i in 0..  
    Dst[i] = Src[Inds[i]];
```

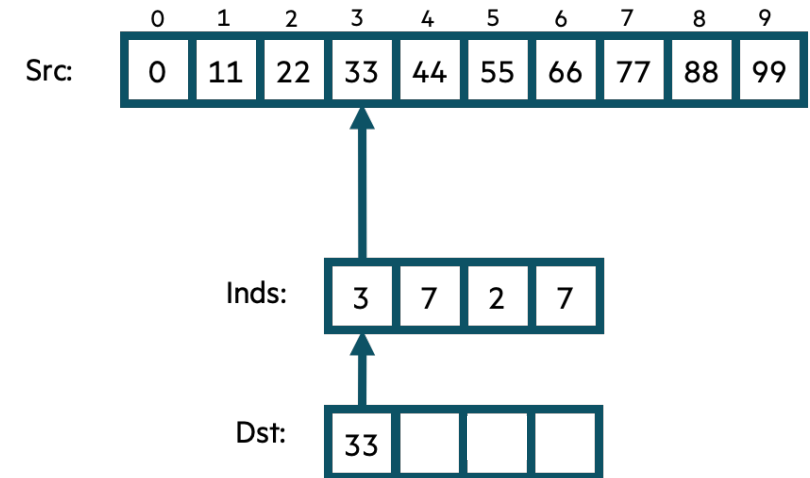
```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```



# Bale IG in Chapel: Serial, Zippered Version

```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    Inds, Dst: [0..  
...  
for (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

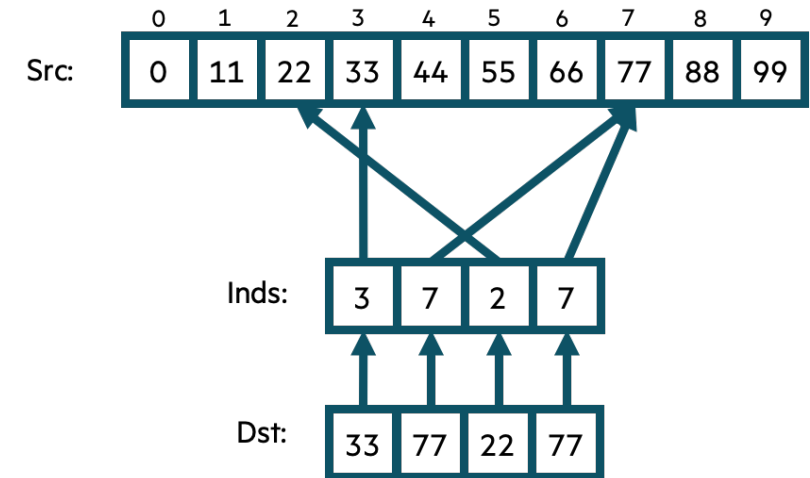
```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```



# Bale IG in Chapel: Parallel, Zippered Version (Vectorized)

```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    n] int,  
    Inds, Dst: [0..  
    m] int;  
...  
foreach (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```

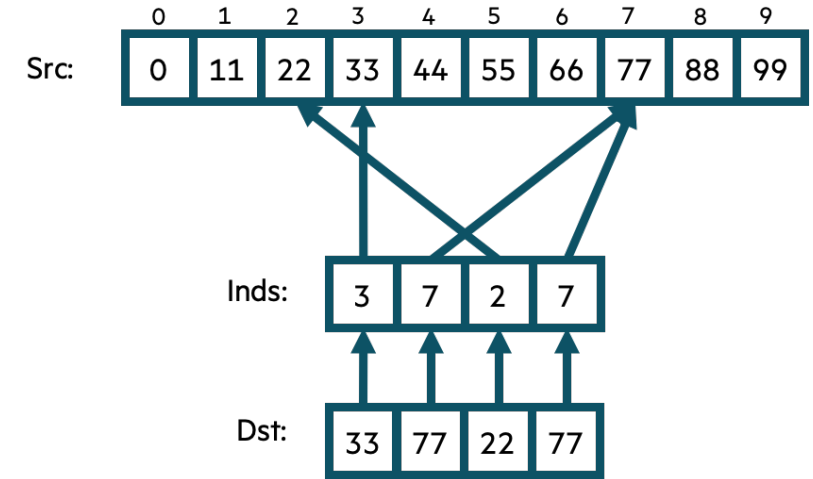




# Bale IG in Chapel: Parallel, Zippered Version (Multicore)

```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    n] int,  
    Inds, Dst: [0..  
    m] int;  
...  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

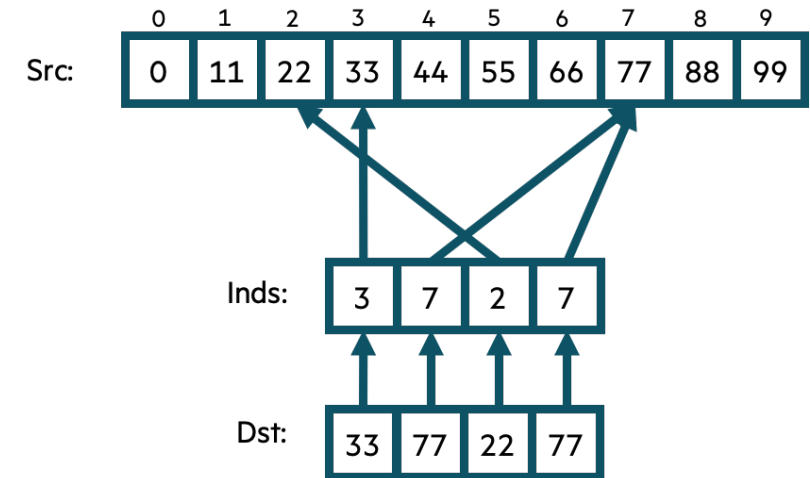
```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```



# Bale IG in Chapel: Parallel Promoted Version (equivalent to previous version)

```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    Inds, Dst: [0..  
...  
Dst = Src[Inds];
```

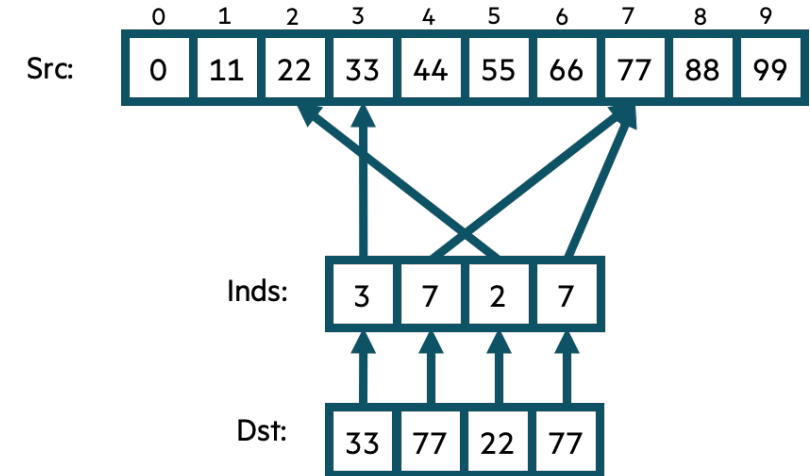
```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```



# Bale IG in Chapel: Parallel, Zippered Version (Multicore)

```
config const n = 10,  
            m = 4;  
  
var Src: [0..  
    Inds, Dst: [0..  
...  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```

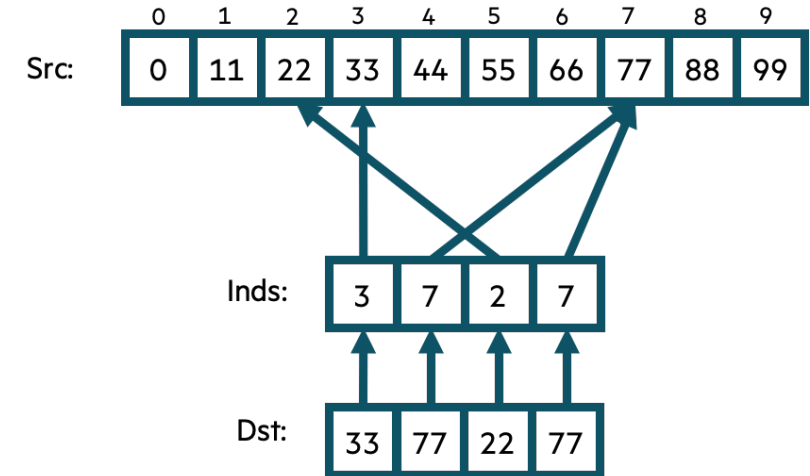


# Bale IG in Chapel: Parallel, Zippered Version for a GPU

```
config const n = 10,  
            m = 4;
```

```
on here.gpus[0] {  
  var Src: [0..<n] int,  
      Inds, Dst: [0..<m] int;  
  
  ...  
  forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];  
}
```

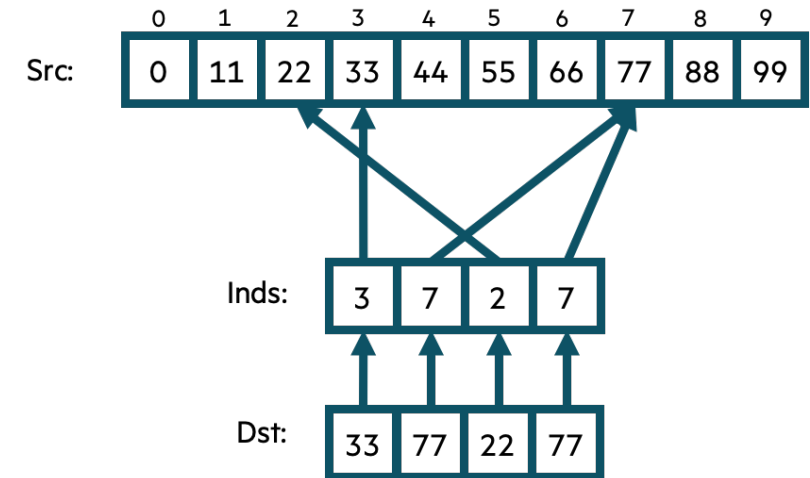
```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```



# Bale IG in Chapel: Parallel, Zippered Version (Multicore)

```
config const n = 10,  
            m = 4;  
  
var Src: [0..<n] int,  
    Inds, Dst: [0..<m] int;  
...  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

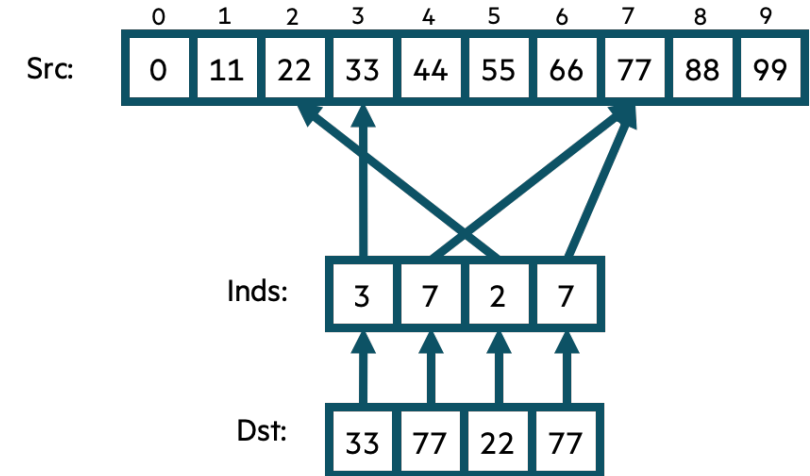
```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```



# Bale IG in Chapel: Parallel , Zippered Version with Named Domains (Multicore)

```
config const n = 10,  
            m = 4;  
  
const SrcInds = {0..  
            DstInds = {0..  
  
var Src: [SrcInds] int,  
     Inds, Dst: [DstInds] int;  
...  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

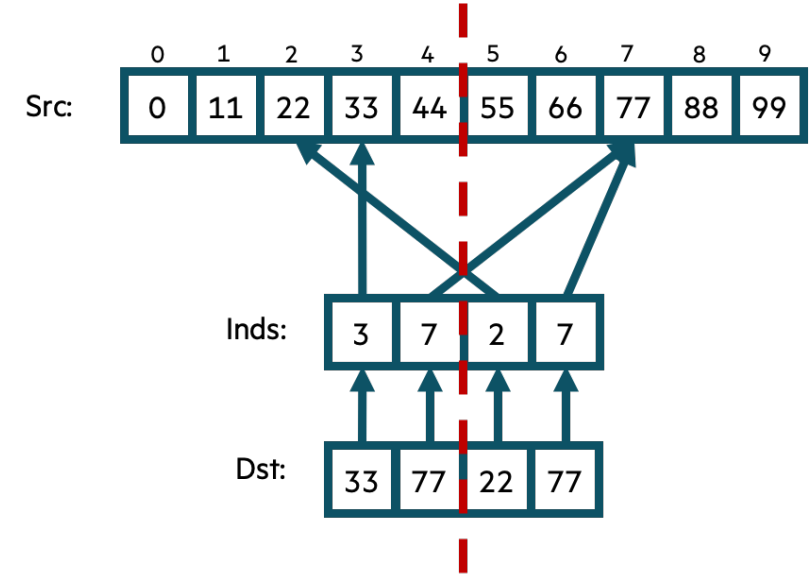
```
$ chpl bale-ig.chpl  
$ ./bale-ig  
$
```



# Bale IG in Chapel: Distributed Parallel Version

```
use BlockDist;  
  
config const n = 10,  
            m = 4;  
  
const SrcInds = blockDist.createDomain(0..  
    DstInds = blockDist.createDomain(0..  
  
var Src: [SrcInds] int,  
     Inds, Dst: [DstInds] int;  
...  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

```
$ chpl bale-ig.chpl  
$ ./bale-ig -nl 4096  
$
```



# Bale IG in Chapel: Distributed Parallel Version

```
use BlockDist;

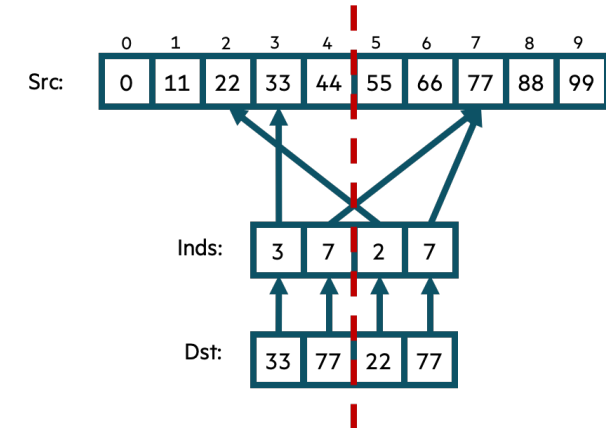
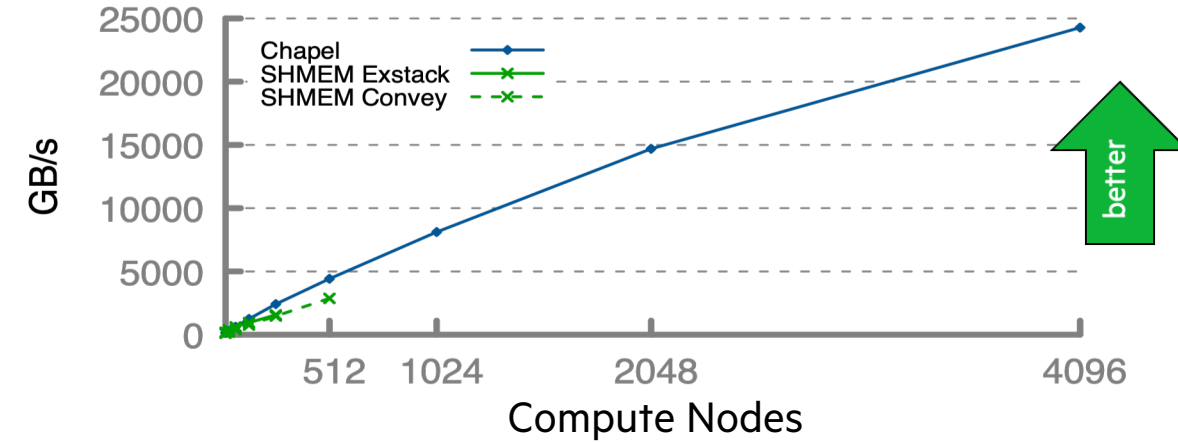
config const n = 10,
             m = 4;

const SrcInds = blockDist.createDomain(0..
```

```
$ chpl bale-ig.chpl --auto-aggregation
$ ./bale-ig -nl 4096
$
```

Bale Indexgather Performance

HPE Cray EX (Slingshot-11)





# Bale IG in Chapel vs. SHMEM on HPE Cray EX (Slingshot-11)

## Chapel (Simple / Auto-Aggregated version)

```
forall (d, i) in zip(Dst, Inds) do
  d = Src[i];
```

## SHMEM (Exstack version)

```
i=0;
while( exstack_proceed(ex, (i==l_num_req)) ) {
  i0 = i;
  while(i < l_num_req) {
    l_indx = pckindx[i] >> 16;
    pe = pckindx[i] & 0xffff;
    if(!exstack_push(ex, &l_indx, pe))
      break;
    i++;
  }

  exstack_exchange(ex);

  while(exstack_pop(ex, &idx, &fromth)) {
    idx = ltable[idx];
    exstack_push(ex, &idx, fromth);
  }
  lgp_barrier();
  exstack_exchange(ex);

  for(j=i0; j<i; j++) {
    fromth = pckindx[j] & 0xffff;
    exstack_pop_thread(ex, &idx, (uint64_t)fromth);
    tgt[j] = idx;
  }
  lgp_barrier();
}
```

## SHMEM (Conveyors version)

```
i = 0;
while (more = convey_advance(requests, (i == l_num_req)),
       more | convey_advance(replies, !more)) {

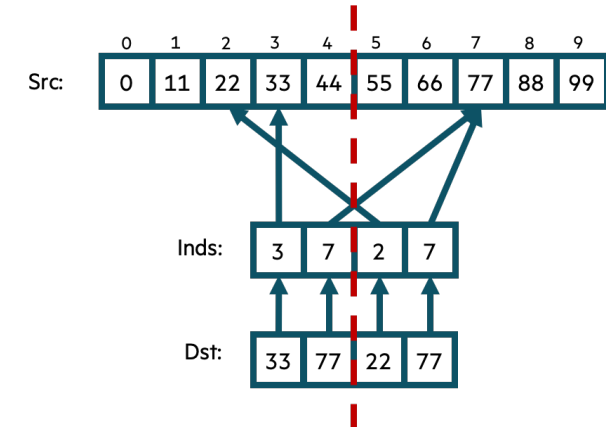
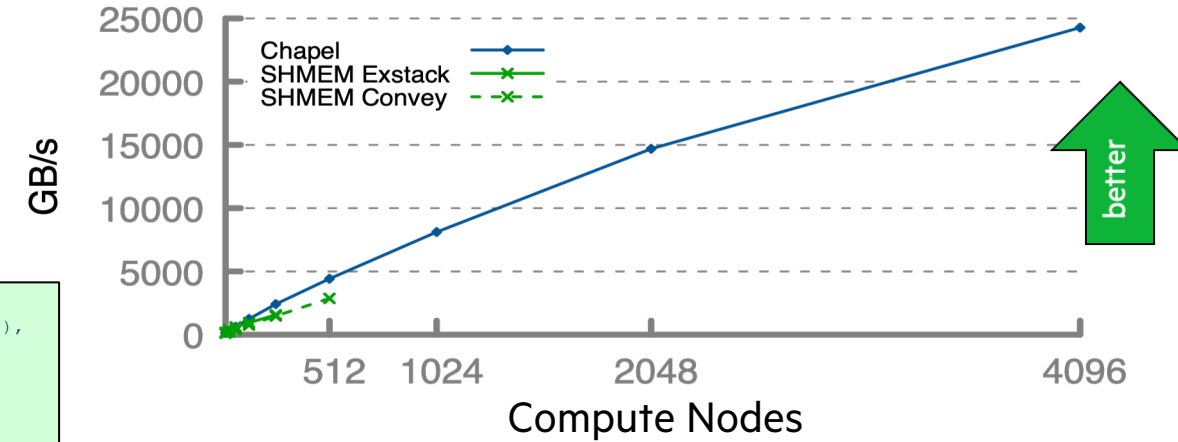
  for (; i < l_num_req; i++) {
    pkg.idx = i;
    pkg.val = pckindx[i] >> 16;
    pe = pckindx[i] & 0xffff;
    if (!convey_push(requests, &pkg, pe))
      break;
  }

  while (convey_pull(requests, ptr, &from) == convey_OK) {
    pkg.idx = ptr->idx;
    pkg.val = ltable[ptr->val];
    if (!convey_push(replies, &pkg, from)) {
      convey_unpull(requests);
      break;
    }
  }

  while (convey_pull(replies, ptr, NULL) == convey_OK)
    tgt[ptr->idx] = ptr->val;
}
```

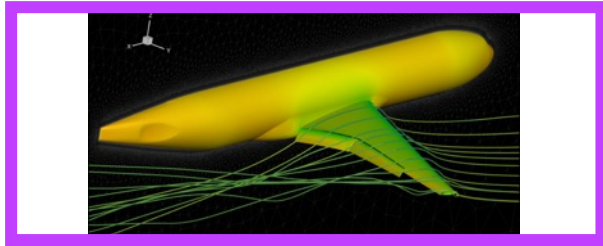
## Bale Indexgather Performance

HPE Cray EX (Slingshot-11)



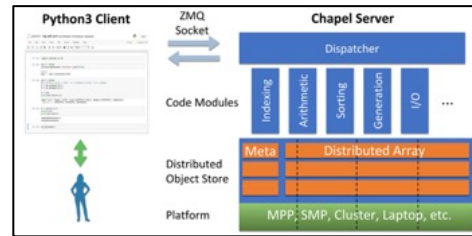
# Applications of Chapel

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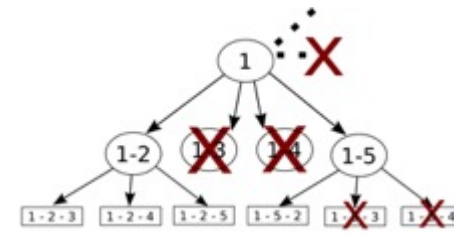
## CHAMPS: 3D Unstructured CFD

Laurendeau, Bourgault-Côté, Parenteau, Plante, et al.  
École Polytechnique Montréal



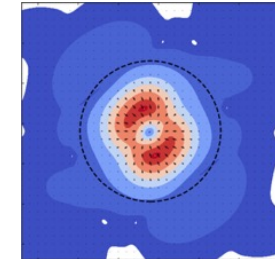
## Arkouda: Interactive Data Science at Massive Scale

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U.S. DoD



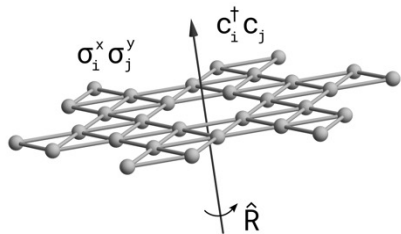
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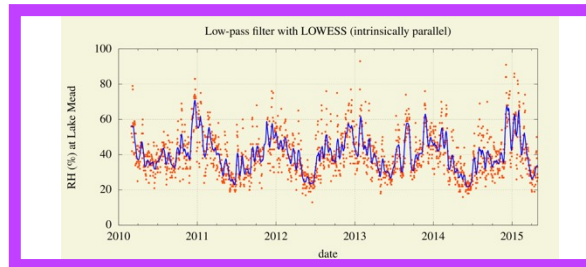
## ChplUltra: Simulating Ultralight Dark Matter

Nikhil Padmanabhan, J. Luna Zagorac, et al.  
Yale University et al.



## Lattice-Symmetries: a Quantum Many-Body Toolbox

Tom Westerhout  
Radboud University



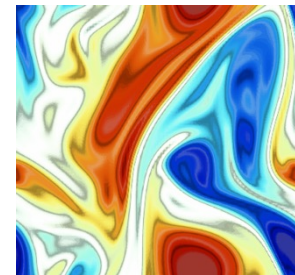
## Desk dot chpl: Utilities for Environmental Eng.

Nelson Luis Dias  
The Federal University of Paraná, Brazil



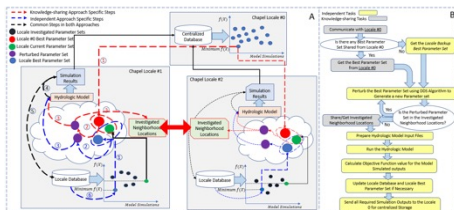
## RapidQ: Mapping Coral Biodiversity

Rebecca Green, Helen Fox, Scott Bachman, et al.  
The Coral Reef Alliance



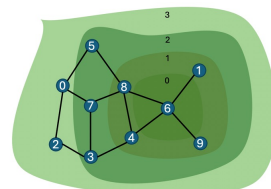
## ChapQG: Layered Quasigeostrophic CFD

Ian Grooms and Scott Bachman  
University of Colorado, Boulder et al.



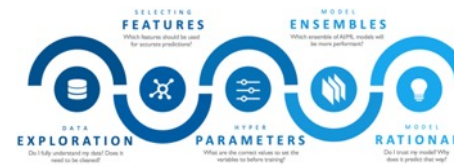
## Chapel-based Hydrological Model Calibration

Marjan Asgari et al.  
University of Guelph



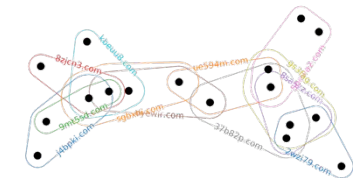
## Arachne Graph Analytics

Bader, Du, Rodriguez, et al.  
New Jersey Institute of Technology



## CrayAI HyperParameter Optimization (HPO)

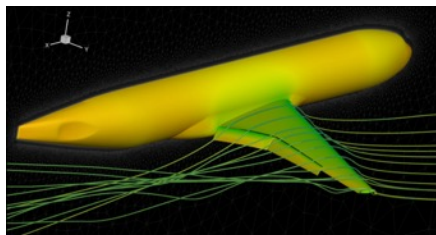
Ben Albrecht et al.  
Cray Inc. / HPE



## CHGL: Chapel Hypergraph Library

Louis Jenkins, Cliff Joslyn, Jesun Firoz, et al.  
PNNL

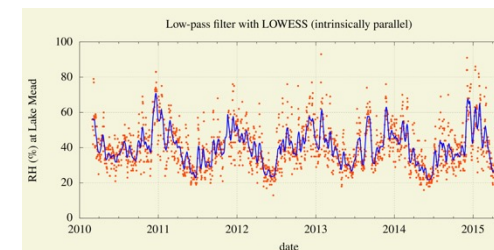
# Productivity Across Diverse Application Scales (code and system size)




**Computation:** Aircraft simulation / CFD  
**Code size:** 100,000+ lines  
**Systems:** Desktops, HPC systems




**Computation:** Coral reef image analysis  
**Code size:** ~300 lines  
**Systems:** Desktops, HPC systems w/ GPUs



**Computation:** Atmospheric data analysis  
**Code size:** 5000+ lines  
**Systems:** Desktops w/ GPUs


 **7 Questions for Éric Laurendeau: Computing Aircraft Aerodynamics in Chapel**  
Posted on September 17, 2024.  
Tags: Computational Fluid Dynamics User Experiences Interviews  
By: Engin Kayraklioglu, Brad Chamberlain

*“Chapel worked as intended: the code maintenance is very much reduced, and its readability is astonishing. This enables undergraduate students to contribute, something almost impossible to think of when using very complex software.”*

 **7 Questions for Scott Bachman: Analyzing Coral Reefs with Chapel**  
Posted on October 1, 2024.  
Tags: Earth Sciences Image Analysis GPU Programming  
User Experiences Interviews  
By: Brad Chamberlain, Engin Kayraklioglu

In this second installment of our [Seven Questions for Chapel Users](#) series, we're looking at a recent success story in which Scott Bachman used Chapel to unlock new scales of biodiversity analysis in coral reefs to study ocean health using satellite image processing. This is work that

*“With the coral reef program, I was able to speed it up by a factor of 10,000. Some of that was algorithmic, but Chapel had the features that allowed me to do it.”*

 **7 Questions for Nelson Luís Dias: Atmospheric Turbulence in Chapel**  
Posted on October 15, 2024.  
Tags: User Experiences Interviews Data Analysis  
Computational Fluid Dynamics  
By: Engin Kayraklioglu, Brad Chamberlain

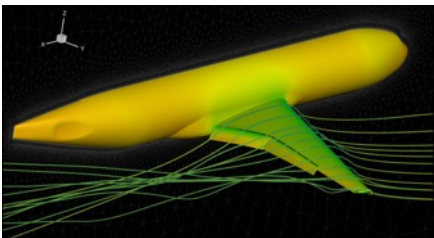
In this edition of our [Seven Questions for Chapel Users](#) series, we turn to Dr. Nelson Luis Dias from Brazil who is using Chapel to analyze data generated by the Amazon Tall Tower Observatory (ATTO), a project dedicated to long-term, 24/7 monitoring of greenhouse gas fluctuations. Read on

*“Chapel allows me to use the available CPU and GPU power efficiently without low-level programming of data synchronization, managing threads, etc.”*

[read this interview series at: <https://chapel-lang.org/blog/series/7-questions-for-chapel-users/>]

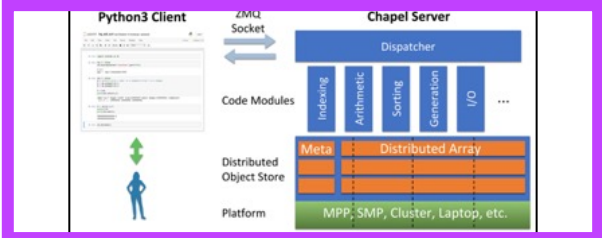


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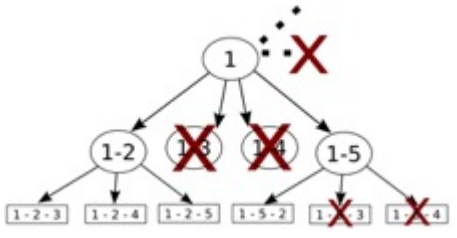
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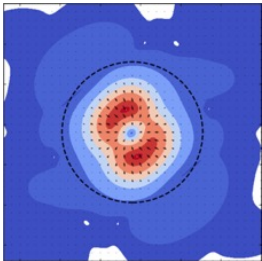
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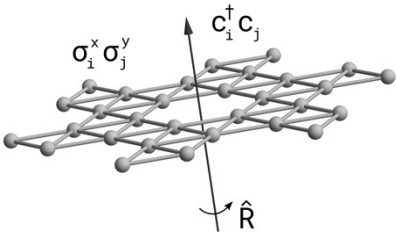
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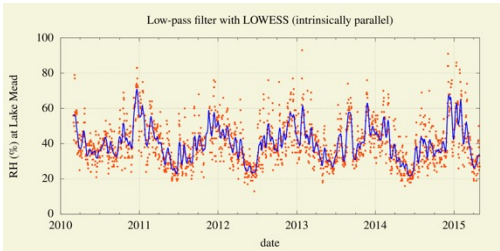
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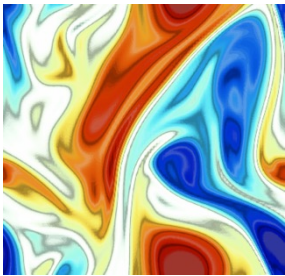
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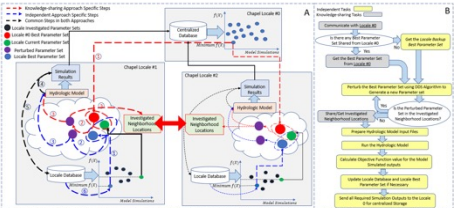
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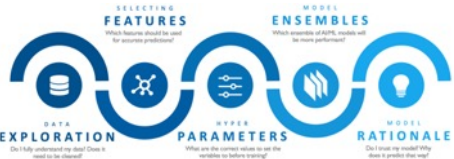
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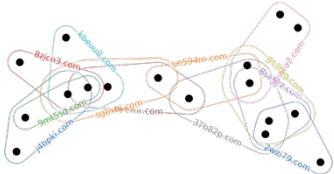
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Ben Albrecht et al.  
*Cray Inc. / HPE*



**CHGL: Chapel Hypergraph Library**  
Louis Jenkins, Cliff Joslyn, Jesun Firoz, et al.  
*PNNL*



**Your Application Here?**

(images provided by their respective teams and used with permission)

# Data Science In Python at scale?

**Motivation:** Imagine you've got...

...HPC-scale data science problems to solve

...a bunch of Python programmers

...access to HPC systems



How will you leverage your Python programmers to get your work done?

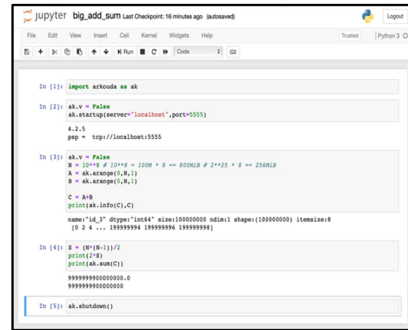


# What is Arkouda?

**Q:** “What is Arkouda?”



**Arkouda Client**  
(written in Python)

A screenshot of a Jupyter Notebook interface. The title bar says 'jupyter big\_add\_sum Last Checkpoint: 10 minutes ago (auto-save)'. The code area contains several lines of Python code that interact with the Arkouda client. The output area shows the results of the code execution, including a large array of numbers.

```
In [1]: import arkouda as ak

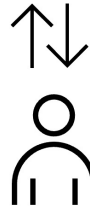
In [2]: ak.v = False
ak.startup(server="localhost", port=5555)
4.2.5
prep = http://localhost:5555

In [3]: ak.v = False
N = 10**8 # 10^8 = 100M * 8 == 800MB # 2**25 = 33M * 8 == 256MB
A = ak.arange(N, N+1)
B = ak.arange(N, N+1)
C = A*B
print(ak.info(C, C))

name: 'A_B' dtype: 'int64' size: 100000000 ndim: 1 shape: (100000000,) itemsize: 8
[0 0 1 ... 39999999 100000000 100000000]

In [4]: S = ak.random(10**7)
print(S)
print(ak.sum(C))
999999900000000.0
999999900000000.0

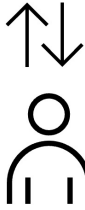
In [5]: ak.shutdown()
```



**User writes Python code**  
**making familiar NumPy/Pandas calls**



\_\_\_\_\_





# Performance and Productivity: Arkouda Argosort

## HPE Cray EX

- Slingshot-11 network (200 Gb/s)
- 8192 compute nodes
- 256 TiB of 8-byte values
- ~8500 GiB/s (~31 seconds)

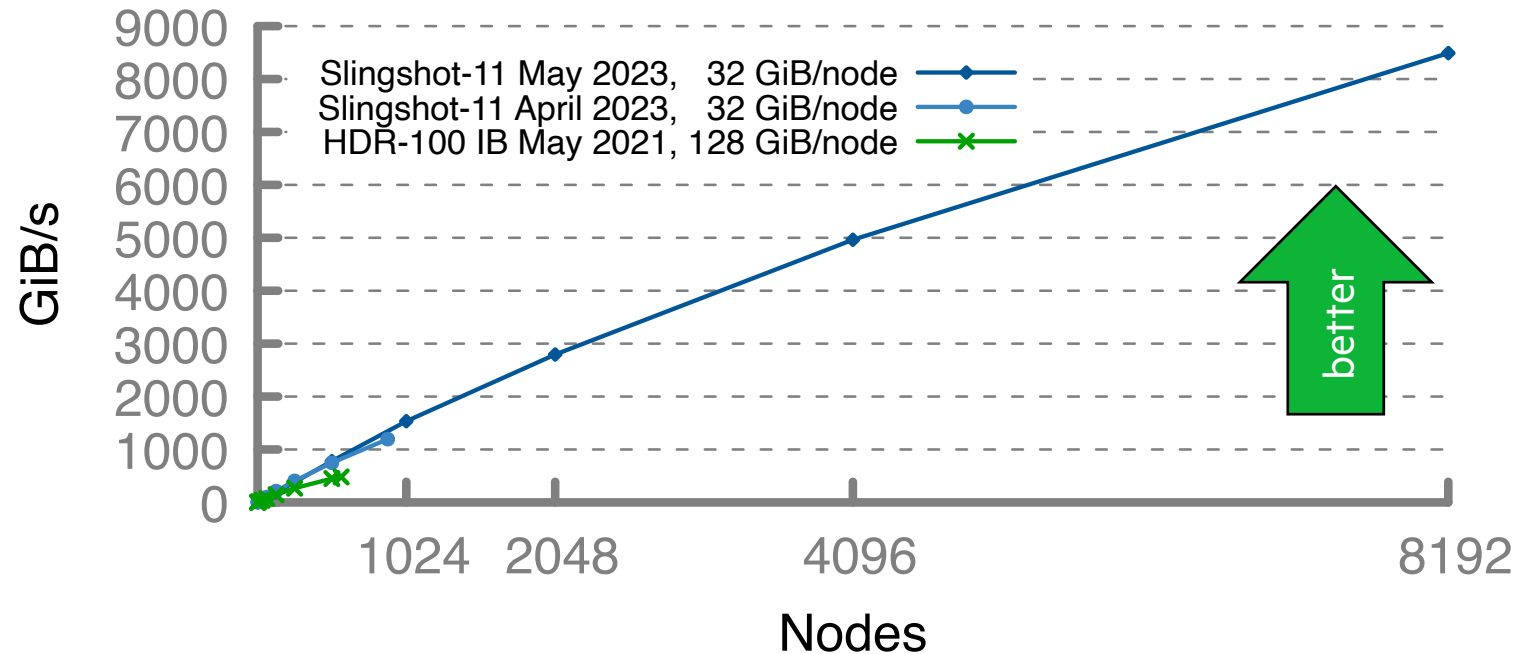
## HPE Cray EX

- Slingshot-11 network (200 Gb/s)
- 896 compute nodes
- 28 TiB of 8-byte values
- ~1200 GiB/s (~24 seconds)

## HPE Apollo

- HDR-100 InfiniBand network (100 Gb/s)
- 576 compute nodes
- 72 TiB of 8-byte values
- ~480 GiB/s (~150 seconds)

Arkouda Argosort Performance



Implemented using ~100 lines of Chapel




\_\_\_\_\_



# Arkouda Resources

**Website:** <https://arkouda-www.github.io/>



Arkouda

[github](#) [documentation](#) [gitter](#)

Massive-scale data science,  
from the comfort of your laptop

Arkouda

Ready for supercomputers

NumPy

Industry standard

```
# Launch an Arkouda server: ./arkouda_server -nl <number-of-locates>
import arkouda as ak

# connect to the server
ak.connect('localhost', 5555)

# Generate two large arrays
a = ak.random.randint(0,2**32,2**30) # ----> Won't fit on a single machine!
b = ak.random.randint(0,2**32,2**30) # 1TB of random integers.

# add them
c = a + b

# Sort the array and print first 10 elements
c = ak.sort(c)
print(c[0:10])
```

Try it Out

Tutorial Video

Chat on Gitter

Arkouda v2024.12.06 released!

The new release includes a refactored server making it easier to add new features, more Sparse Matrix functionality, new pddarray manipulation functions, and bug fixes.

[Read the release notes -->](#)

Arkouda is...

Fast

Arkouda is powered by Chapel, a programming language built from the ground up to support parallelism and distributed computing. Make the most out of every core and every node in your system.

Interactive

By distributing your data across multiple nodes, Arkouda allows you to rapidly transform and wrangle datasets in real time that are simply intractable for a laptop or desktop.


Extensible

One can expand on Arkouda's capabilities, thus enabling arbitrary scalable computations to be performed from Python.

Powered by Chapel

Arkouda's backend is implemented in Chapel, an open-source parallel programming language. Chapel is unique among mainstream languages as it puts parallelism and locality in the forefront, while not sacrificing productivity or portability. Chapel enables Arkouda to perform well and scale on many different architectures, from multicore laptops to cloud systems to world's fastest supercomputers.

To learn more about Chapel, check out its [blog](#), [presentations](#), [tutorials](#) and [demos](#), and the [How Can I Learn Chapel?](#) page.



Arkouda users are saying...

"

...solving problems in a matter of seconds, as opposed to days...

— Tess Hayes, Bytoia

"


['I'm] working with more data than I ever thought possible as a data scientist

— Jake Trookman, Erias

**GitHub:** <https://github.com/Bears-R-Us/arkouda>

README

License



Arkouda (αρκούδα)

Interactive Data Analytics at Supercomputing Scale

CI passing

docs passing

license MIT

code style black


Online Documentation

[Arkouda docs at Github Pages](#)

Nightly Arkouda Performance Charts

[Arkouda nightly performance charts](#)

**Coming Soon:** interview with founding dev, Bill Reus



7 Questions for Bill Reus: Interactive, Massive-Scale Data Analytics in Chapel

Posted on January 15, 2025.

Tags:

User Experiences

Interviews

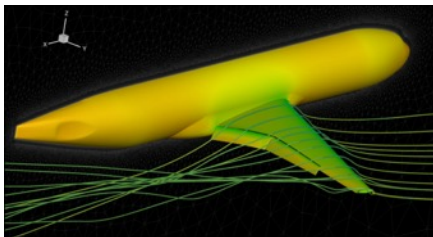
Data Analysis

Arkouda

By: [Engin Kayraklioglu](#), [Brad Chamberlain](#)

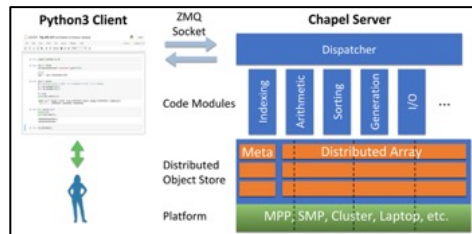
35

# Applications of Chapel: Links to Users' Talks (slides + video) & Blog Interviews



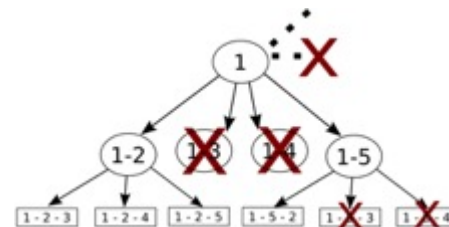
CHAMPS: 3D Unstructured CFD

[CHIOW 2021](#) [CHIOW 2022](#) [Blog](#)



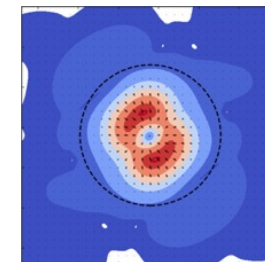
Arkouda: Interactive Data Science at Massive Scale

[CHIOW 2020](#) [CHIOW 2023](#)



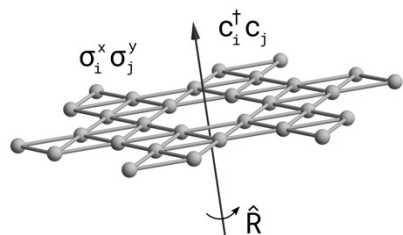
ChOp: Chapel-based Optimization

[CHIOW 2021](#) [2023](#) [ChapelCon'24 \(IJ\)](#)



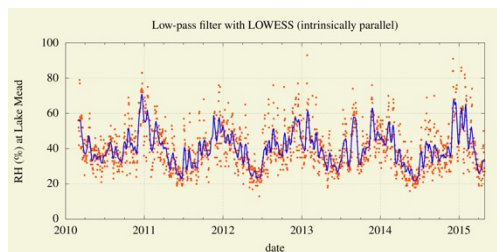
CholUltra: Simulating Ultralight Dark Matter

[CHIOW 2020](#) [CHIOW 2022](#)



Lattice-Symmetries: a Quantum Many-Body Toolbox Desk dot chpl: Utilities for Environmental Eng.

[CHIOW 2022](#)

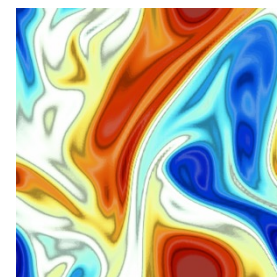


[CHIOW 2022](#) [ChapelCon '24](#) [Blog](#)

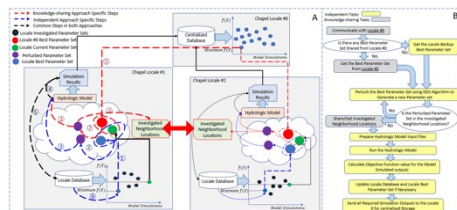


RapidQ: Mapping Coral Biodiversity

[CHIOW 2023](#) [Blog](#)

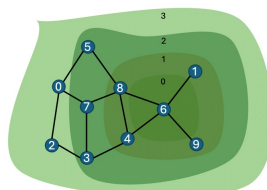


ChanQG: Layered Quasigeostrophic CFD



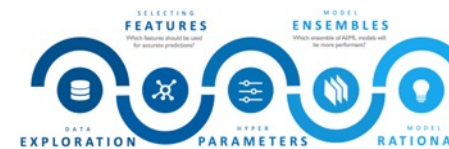
Chapel-based Hydrological Model Calibration

[CHIOW 2023](#)



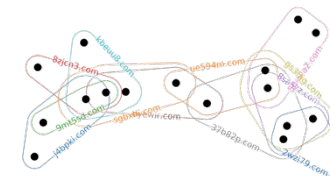
Arachne Graph Analytics

[CHIOW 2023](#) [ChapelCon '24](#) [Blog](#)



CrayAI HyperParameter Optimization (HPO)

[CHIOW 2021](#)



CHGL: Chapel Hypergraph Library

[CHIOW 2020](#)

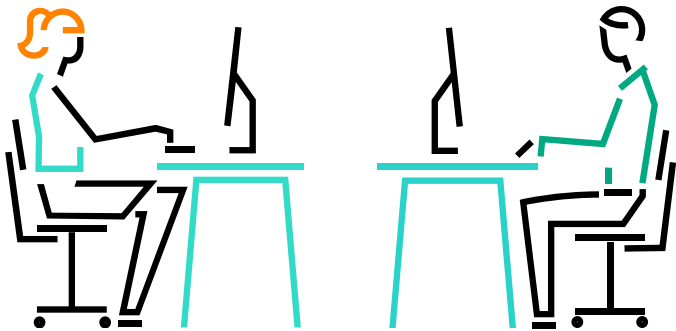
**[NOTE: This slide focuses on presentations and blogs published in Chapel venues, but numerous external publications also exist]**

(images provided by their respective teams and used with permission)

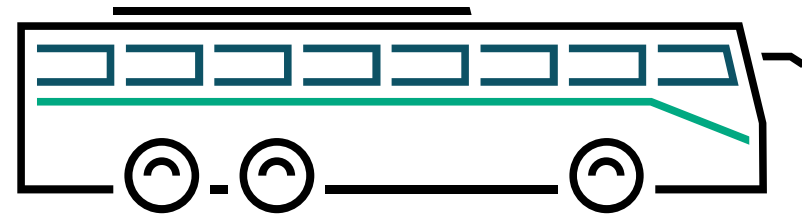
# Global-view vs. SPMD Programming

## A Strained(?) Analogy

Gosh, I bet those  
supercomputer users  
have some swanky  
programming languages...



Gosh, those Le Mans racers  
must have an enviable  
driving experience...





# HPC Benchmarks Using Conventional Programming Approaches

## STREAM TRIAD: C + MPI + OPENMP

```
#include <hpcc.h>
#ifdef _OPENMP
#include <omp.h>
#endif

static int VectorSize;
static double *a, *b, *c;

int HPCC_StarStream(HPCC_Params *params) {
    int myRank, commSize;
    int rv, errCount;
    MPI_Comm comm = MPI_COMM_WORLD;

    MPI_Comm_size( comm, &commSize );
    MPI_Comm_rank( comm, &myRank );

    rv = HPCC_Stream( params, 0 == myRank );
    MPI_Reduce( &rv, &errCount, 1, MPI_INT, MPI_SUM, 0, comm );

    return errCount;
}

int HPCC_Stream(HPCC_Params *params, int doIO) {
    register int j;
    double scalar;

    VectorSize = HPCC_LocalVectorSize( params, 3, sizeof(double), 0 );

    a = HPCC_XMALLOC( double, VectorSize );
    b = HPCC_XMALLOC( double, VectorSize );
    c = HPCC_XMALLOC( double, VectorSize );

    if (!a || !b || !c) {
        if (c) HPCC_free(c);
        if (b) HPCC_free(b);
        if (a) HPCC_free(a);
        if (doIO) {
            fprintf( outFile, "Failed to allocate memory (%d).\n", VectorSize );
            fclose( outFile );
        }
        return 1;
    }

#ifdef _OPENMP
#pragma omp parallel for
#endif
    for (j=0; j<VectorSize; j++) {
        b[j] = 2.0;
        c[j] = 1.0;
    }
    scalar = 3.0;

#ifdef _OPENMP
#pragma omp parallel for
#endif
    for (j=0; j<VectorSize; j++) {
        a[j] = b[j]+scalar*c[j];
    }

    HPCC_free(c);
    HPCC_free(b);
    HPCC_free(a);

    return 0;
}
```

## HPCC RA: MPI KERNEL

```
/* Perform updates to main table. The scalar equivalent is:
 * for (i=0; i<NUPDATE; i++) {
 *   Ran = (Ran << 1) ^ ((s64int) Ran < 0) ? POLY: 0;
 *   Table[Ran & (TABLESIZE-1)] ^= Ran;
 * }
 */

MPI_Irecv(&LocalRecvBuffer, localBufferSize, tparams.dtype64,
          MPI_ANY_SOURCE, MPI_ANY_TAG, MPI_COMM_WORLD, &inreq);
while (! < SendCnt) {
    /* receive messages */
    do {
        MPI_Test(&inreq, &have_done, &status);
        if (have_done) {
            if (status.MPI_TAG == UPDATE_TAG) {
                MPI_Get_count(&status, tparams.dtype64, &recvUpdates);
                bufferBase = 0;
                for (j=0; j < recvUpdates; j++) {
                    inmsg = LocalRecvBuffer[bufferBase+j];
                    LocalOffset = (inmsg & (tparams.TableSize - 1)) -
                        tparams.GlobalStartMyProc;
                    HPCC_Table[LocalOffset] ^= inmsg;
                }
            } else if (status.MPI_TAG == FINISHED_TAG) {
                NumberReceiving--;
            } else {
                MPI_Abort( MPI_COMM_WORLD, -1 );
            }
            MPI_Irecv(&LocalRecvBuffer, localBufferSize, tparams.dtype64,
                    MPI_ANY_SOURCE, MPI_ANY_TAG, MPI_COMM_WORLD, &inreq);
        }
    } while (have_done && NumberReceiving > 0);
    if (pendingUpdates < maxPendingUpdates) {
        Ran = (Ran << 1) ^ ((s64int) Ran < ZERO64B ? POLY : ZERO64B);
        GlobalOffset = Ran & (tparams.TableSize-1);
        if (GlobalOffset < tparams.Top)
            WhichPe = (GlobalOffset / (tparams.MinLocalTableSize + 1));
        else
            WhichPe = ((GlobalOffset - tparams.Remainder) /
                        tparams.MinLocalTableSize);
        if (WhichPe == tparams.MyProc) {
            LocalOffset = (Ran & (tparams.TableSize - 1)) -
                tparams.GlobalStartMyProc;
            HPCC_Table[LocalOffset] ^= Ran;
        }
    } else {
        HPCC_InsertUpdate(Ran, WhichPe, Buckets);
        pendingUpdates++;
    }
    i++;
}
else {
    MPI_Test(&outreq, &have_done, MPI_STATUS_IGNORE);
    if (have_done) {
        outreq = MPI_REQUEST_NULL;
        pe = HPCC_GetUpdates(Buckets, LocalSendBuffer, localBufferSize,
                             &peUpdates);
        MPI_Isend(&LocalSendBuffer, peUpdates, tparams.dtype64, (int)pe,
                  UPDATE_TAG, MPI_COMM_WORLD, &outreq);
        pendingUpdates -= peUpdates;
    }
}

/* send our done messages */
for (proc_count = 0; proc_count < tparams.NumProcs; ++proc_count) {
    if (proc_count == tparams.MyProc) { tparams.finish_req[tparams.MyProc] =
        MPI_REQUEST_NULL; continue; }
    /* send garbage - who cares, no one will look at it */
    MPI_Isend(&Ran, 0, tparams.dtype64, proc_count, FINISHED_TAG,
              MPI_COMM_WORLD, tparams.finish_req + proc_count);
}

/* Finish everyone else up... */
while (NumberReceiving > 0) {
    MPI_Wait(&inreq, &status);
    if (status.MPI_TAG == UPDATE_TAG) {
        MPI_Get_count(&status, tparams.dtype64, &recvUpdates);
        bufferBase = 0;
        for (j=0; j < recvUpdates; j++) {
            inmsg = LocalRecvBuffer[bufferBase+j];
            LocalOffset = (inmsg & (tparams.TableSize - 1)) -
                tparams.GlobalStartMyProc;
            HPCC_Table[LocalOffset] ^= inmsg;
        }
    } else if (status.MPI_TAG == FINISHED_TAG) {
        /* we got a done message. Thanks for playing. */
        NumberReceiving--;
    } else {
        MPI_Abort( MPI_COMM_WORLD, -1 );
    }
    MPI_Irecv(&LocalRecvBuffer, localBufferSize, tparams.dtype64,
              MPI_ANY_SOURCE, MPI_ANY_TAG, MPI_COMM_WORLD, &inreq);
}

MPI_Waitall( tparams.NumProcs, tparams.finish_req, tparams.finish_statuses);
```

Page 10

```
#include <hpc++.h>
#ifdef _OPENMP
#include <omp.h>
#endif

static int VectorSize;
static double *a, *b, *c;

int HPCC_StartStream(HPCC_Params *params) {
    int myRank, commSize;
    int rv, errCount;
    MPI_Comm comm = MPI_COMM_WORLD;

    MPI_Comm_size(comm, &commSize);
    MPI_Comm_rank(comm, &myRank);

    rv = HPCC_Stream(comm, 0 == myRank);
    MPI_Reduce(&rv, &errCount, 1, MPI_INT, MPI_SUM, 0, comm);

    return errCount;
}

int HPCC_Stream(HPCC_Params *params, int doIO) {
    register int j;
    double scalar;

    VectorSize = HPCC_LocalVectorSize(params, 3, sizeof(double), 0);

    a = HPCC_XMALLOC(double, VectorSize);
    b = HPCC_XMALLOC(double, VectorSize);
    c = HPCC_XMALLOC(double, VectorSize);
}
```

```
use BlockDist;
```

```
config const n = 1_000_000,  
            alpha = 0.01;  
const Dom = blockDist.createDomain({1..n});  
var A, B, C: [Dom] real;
```

```
B = 2.0;  
C = 1.0;
```

```
A = B + alpha * C;
```

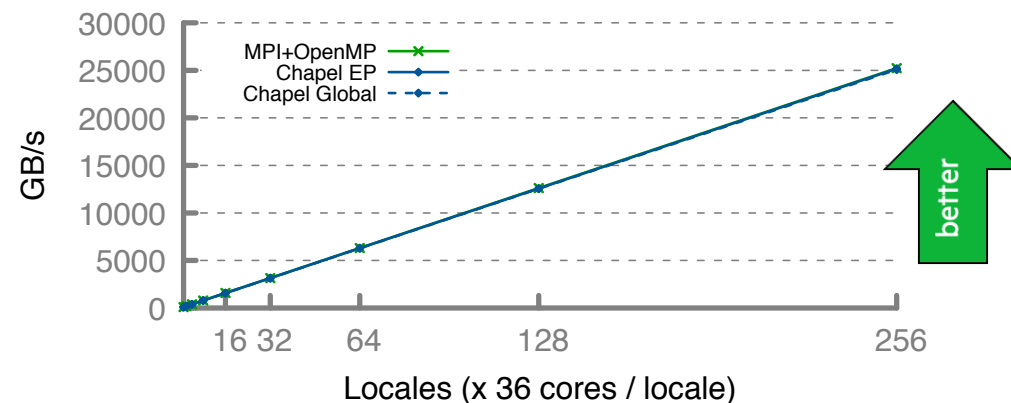
110

[illegible]

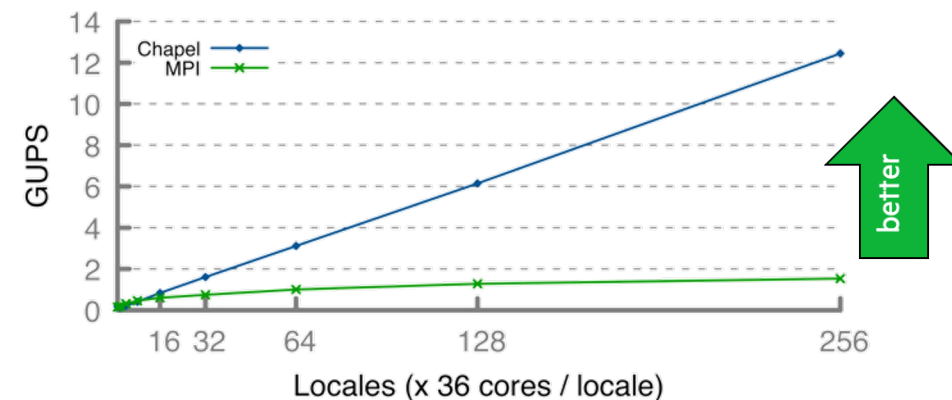
```
...
forall (_, r) in zip(Updates, RASStream()) do
    T[r & indexMask].xor(r);
...
```

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### STREAM Performance (GB/s)



RA Performance (GUPS)





# Bale IG in Chapel vs. SHMEM on HPE Cray EX (Slingshot-11)

## Chapel (Simple / Auto-Aggregated version)

```
forall (d, i) in zip(Dst, Inds) do
  d = Src[i];
```

## SHMEM (Exstack version)

```
i=0;
while( exstack_proceed(ex, (i==l_num_req)) ) {
  i0 = i;
  while(i < l_num_req) {
    l_indx = pckindx[i] >> 16;
    pe = pckindx[i] & 0xffff;
    if(!exstack_push(ex, &l_indx, pe))
      break;
    i++;
  }

  exstack_exchange(ex);

  while(exstack_pop(ex, &idx, &fromth)) {
    idx = ltable[idx];
    exstack_push(ex, &idx, fromth);
  }
  lgp_barrier();
  exstack_exchange(ex);

  for(j=i0; j<i; j++) {
    fromth = pckindx[j] & 0xffff;
    exstack_pop_thread(ex, &idx, (uint64_t)fromth);
    tgt[j] = idx;
  }
  lgp_barrier();
}
```

## SHMEM (Conveyors version)

```
i = 0;
while (more = convey_advance(requests, (i == l_num_req)),
       more | convey_advance(replies, !more)) {

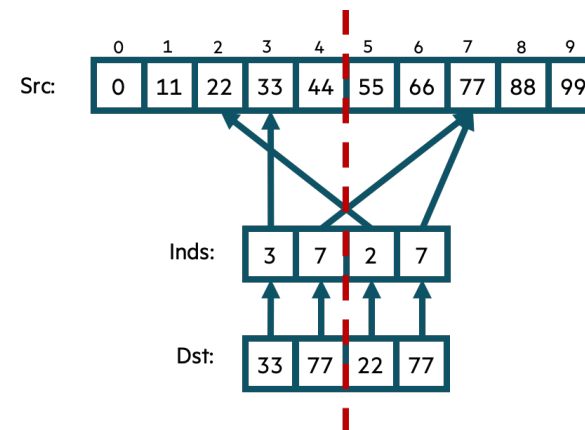
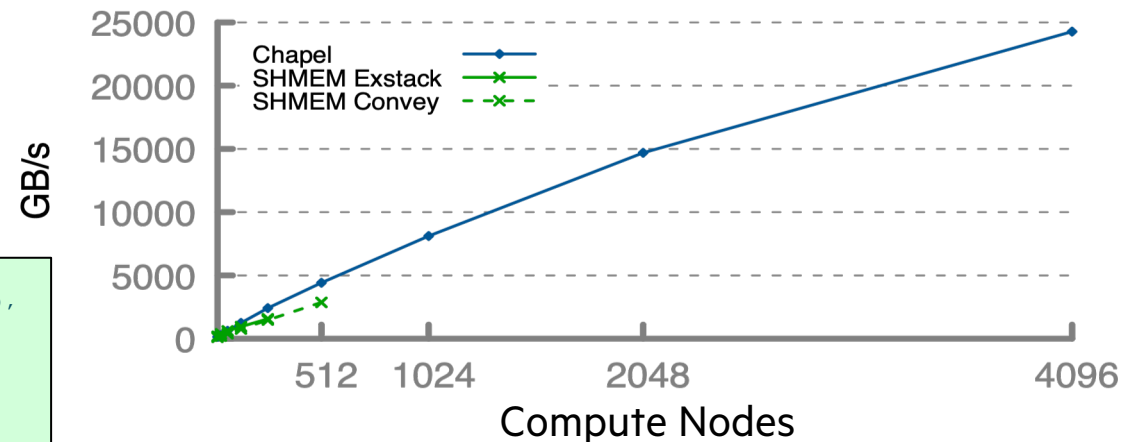
  for (; i < l_num_req; i++) {
    pkg.idx = i;
    pkg.val = pckindx[i] >> 16;
    pe = pckindx[i] & 0xffff;
    if (!convey_push(requests, &pkg, pe))
      break;
  }

  while (convey_pull(requests, ptr, &from) == convey_OK) {
    pkg.idx = ptr->idx;
    pkg.val = ltable[ptr->val];
    if (!convey_push(replies, &pkg, from)) {
      convey_unpull(requests);
      break;
    }
  }

  while (convey_pull(replies, ptr, NULL) == convey_OK)
    tgt[ptr->idx] = ptr->val;
}
```

## Bale Indexgather Performance

HPE Cray EX (Slingshot-11)

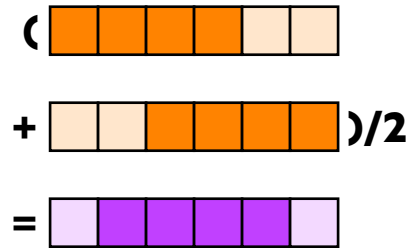


**Q: What accounts for the code size disparities between Chapel and SHMEM / MPI?**

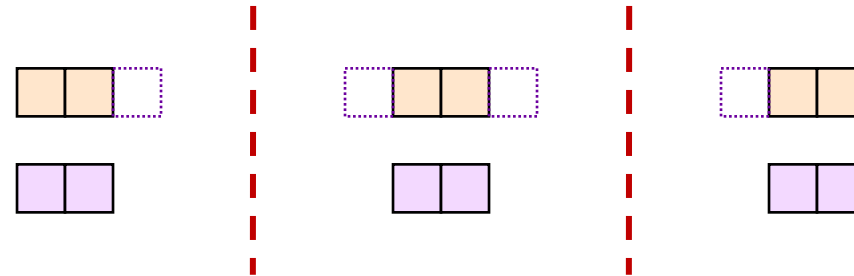
# A: Chapel Supports Global-view Programming

**Example:** “Apply a 3-point stencil to a vector”

*Global-View*



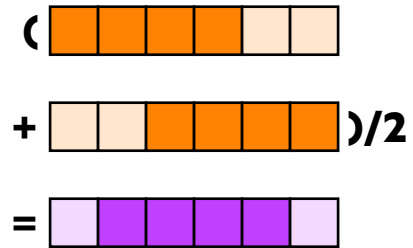
*SPMD*



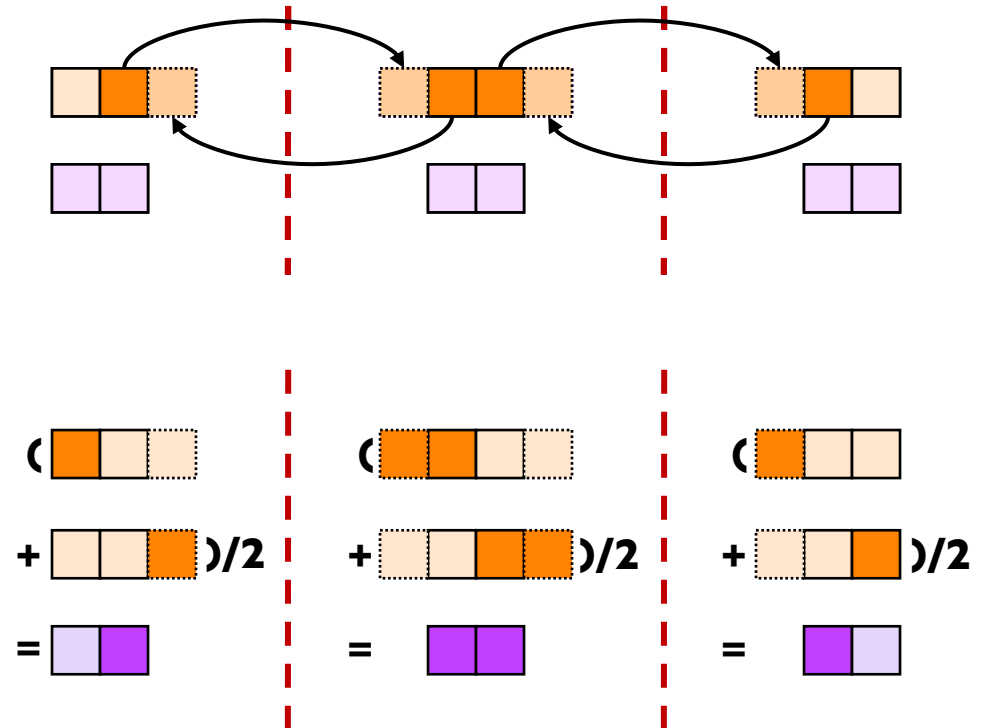
# A: Chapel Supports Global-view Programming

**Example:** “Apply a 3-point stencil to a vector”

*Global-View*




*SPMD*



# A: Chapel Supports Global-view Programming

**Example:** “Apply a 3-point stencil to a vector”

## Global-View Chapel code

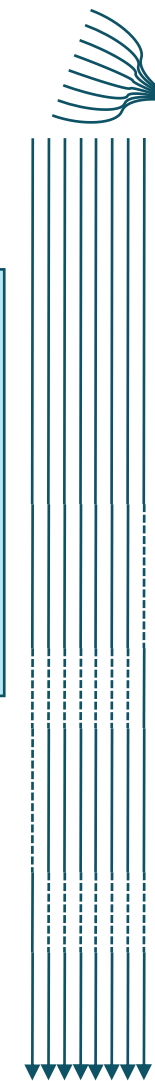


```
use BlockDist;

proc main() {
  var n = 1000;
  const D = blockDist.createDomain(1..n);

  forall i in D[2..n-1] do
    B[i] = (A[i-1] + A[i+1])/2;
  }
}
```

## SPMD pseudocode (MPI-esque)



```
proc main() {
  var n = 1000;
  var p = numProcs(),
      me = myProc(),
      myN = n/p,
      myLo = 1,
      myHi = myN;
  var A, B: [0..myN+1] real;

  if (me < p-1) {
    send(me+1, A[myN]);
    recv(me+1, A[myN+1]);
  } else
    myHi = myN-1;
  if (me > 0) {
    send(me-1, A[1]);
    recv(me-1, A[0]);
  } else
    myLo = 2;
  forall i in myLo..myHi do
    B[i] = (A[i-1] + A[i+1])/2;
  }
}
```

# SPMD Programming in Chapel

That said, as a general-purpose language, Chapel supports writing SPMD patterns as well:

```
coforall loc in Locales do
  on loc do
    myMain();

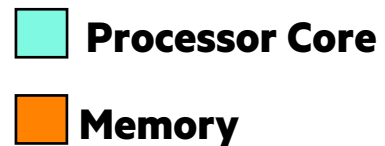
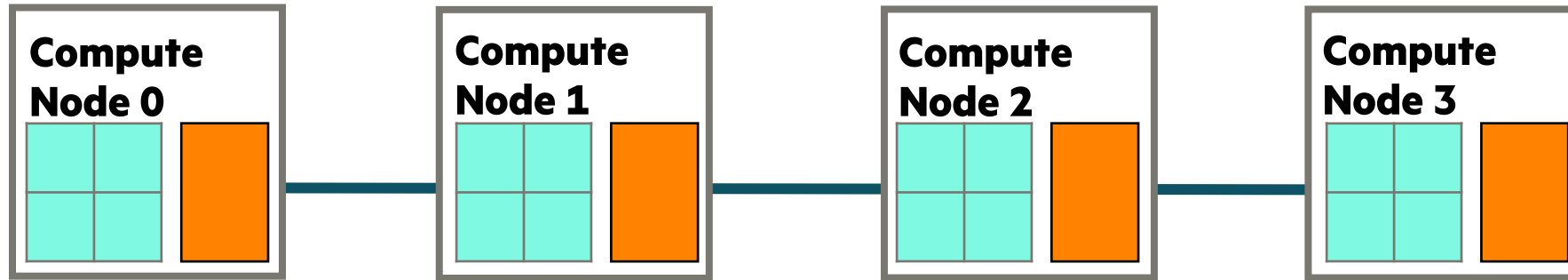
proc myMain() {
  // ... write your SPMD computation here ...
}
```



# **Chapel Features for Parallelism and Locality**

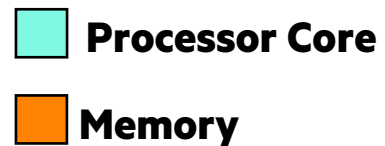
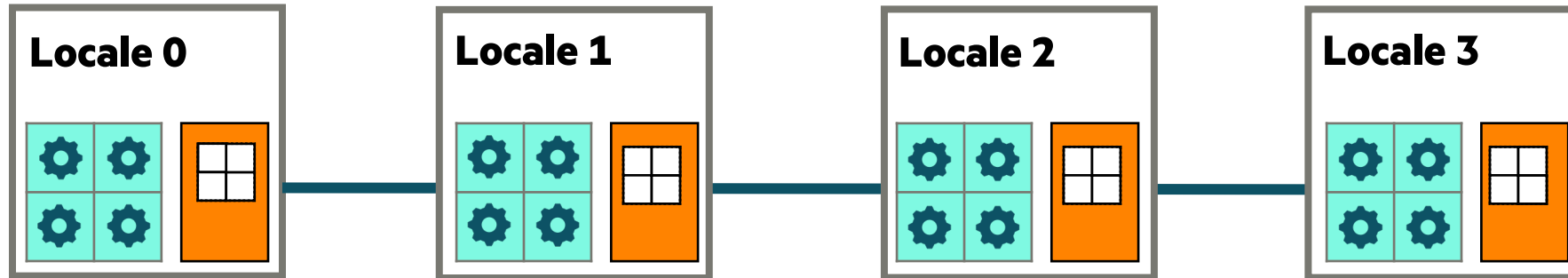
# Locales in Chapel

- In Chapel, a *locale* refers to a compute resource with...
  - processors, so it can run tasks
  - memory, so it can store variables
- For now, think of each compute node as being a locale



# Key Concerns for Scalable Parallel Computing

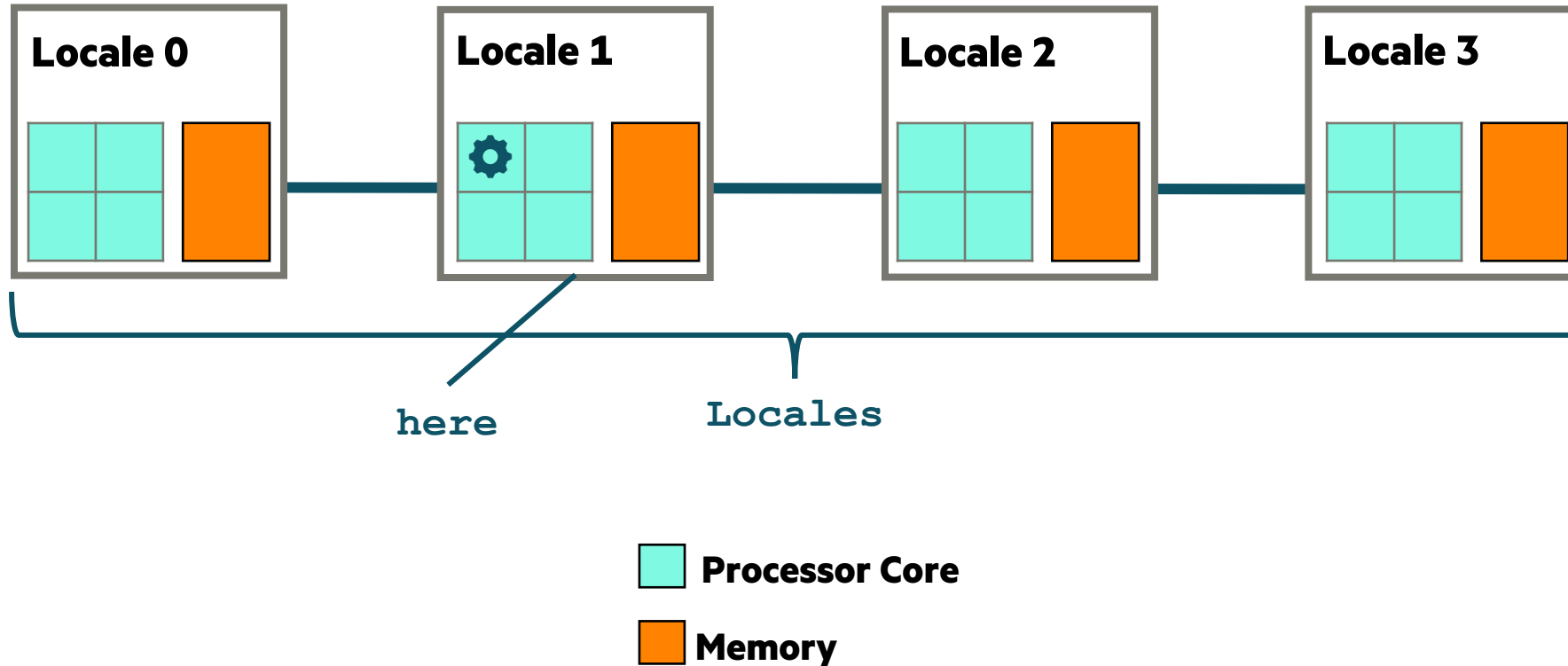
1. **parallelism:** What tasks should run simultaneously?
2. **locality:** Where should tasks run? Where should data be allocated?





# Built-In Locale Variables in Chapel

- Two key built-in variables for referring to locales in Chapel programs:
  - **Locales**: An array of locale values representing the system resources on which the program is running
  - **here**: The locale on which the current task is executing



# Basic Features for Locality

basics-on.chpl

```
writeln("Hello from locale ", here.id);  
  
var A: [1..2, 1..2] real;  
  
on Locales[1] {  
    var B: [1..2, 1..2] real;  
  
    B = 2 * A;  
}
```

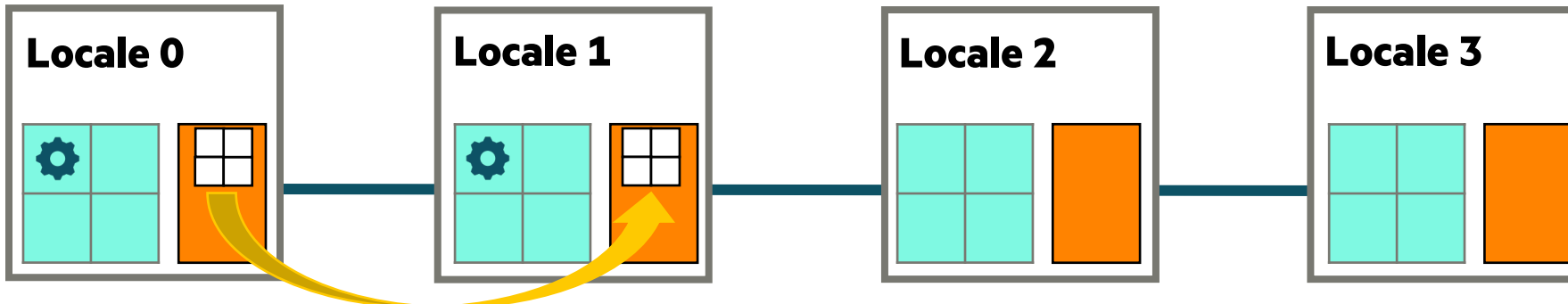
All Chapel programs begin running as a single task on locale 0

Variables are stored using the memory local to the current task

on-clauses move tasks to other locales

remote variables can be accessed directly

**This is a serial, but distributed computation**



# Basic Features for Locality

basics-for.chpl

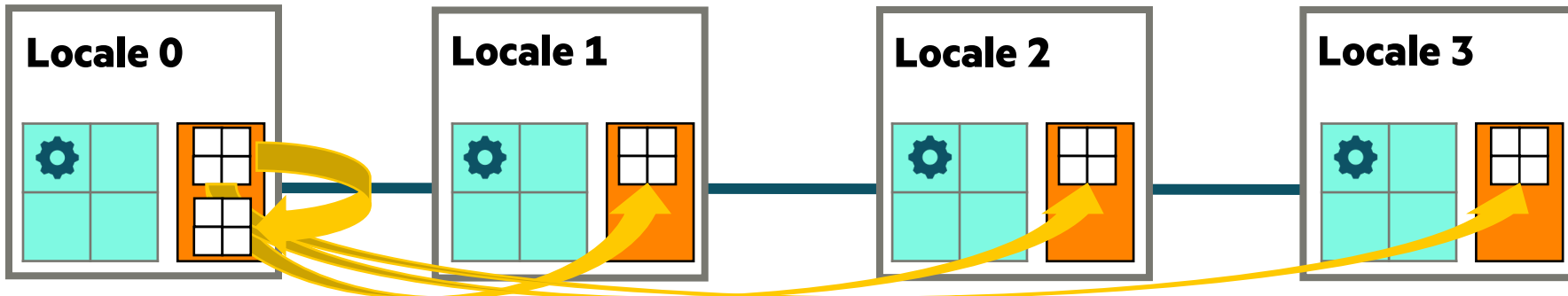
```
writeln("Hello from locale ", here.id);

var A: [1..2, 1..2] real;

for loc in Locales {
  on loc {
    var B = A;
  }
}
```

This loop will serially iterate over the program's locales

This is also a serial, but distributed computation



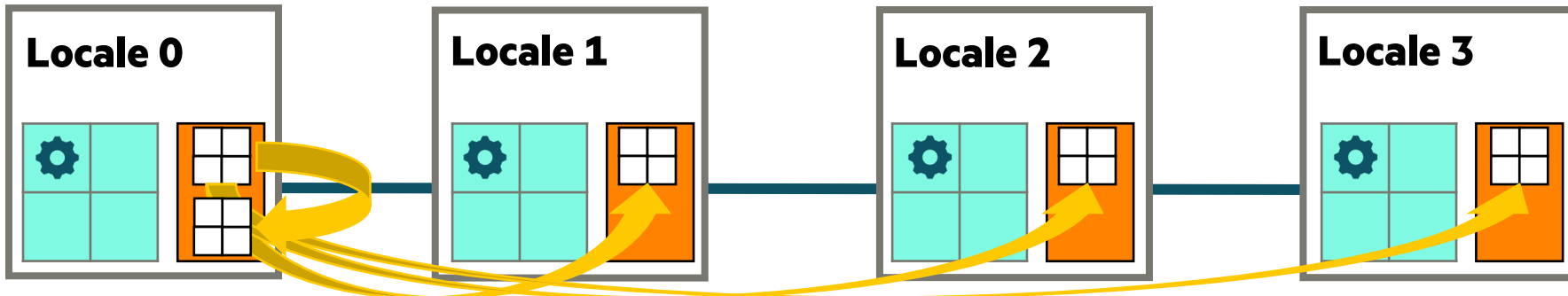
# Mixing Locality with Task Parallelism

basics-coforall.chpl

```
writeln("Hello from locale ", here.id);  
  
var A: [1..2, 1..2] real;  
  
coforall loc in Locales {  
  on loc {  
    var B = A;  
  }  
}
```

The coforall loop creates  
a parallel task per iteration

This results in a parallel distributed computation



# The Three Ways to Create Parallel Tasks in Chapel

**begin:** Creates a task to asynchronously execute the statement it prefixes

```
begin writeln("Hello, PLSE!");  
writeln("Goodbye!");
```

**cobegin:** A compound statement in which each child statement is a distinct task

```
cobegin {  
    writeln("Hello from task 1");  
    writeln("Hello from task 2");  
}  
writeln("Goodbye!"); // original task waits for child tasks to complete before proceeding
```

**coforall:** A loop form in which each iteration is a distinct task

```
coforall i in 1..numTasks do  
    writeln("Hello from task ", i, " of ", numTasks);  
writeln("Goodbye!"); // original task waits for child tasks to complete before proceeding
```



# Wait, what about ‘forall’ and ‘foreach’?

**forall:** Invokes a parallel iterator, itself written in terms of ‘coforall’, ‘cobegin’, and/or ‘begin’

```
forall i in 1..n do
  writeln("Hello from iteration ", i, " of ", n);
writeln("Goodbye!");
```

*// notionally, the parallel iterator for a range looks something like this:*

```
proc range.these(...) {
  const numTasks = computeNumTasks();
  coforall i in 0..<numTasks {
    const chunk = computeMyChunk(lo, hi, stride, numTasks);
    for j in chunk do
      yield j;
    }
  }
}
```

**foreach:** Doesn’t introduce any tasks, just hints to the compiler that the loop may / should be parallelized



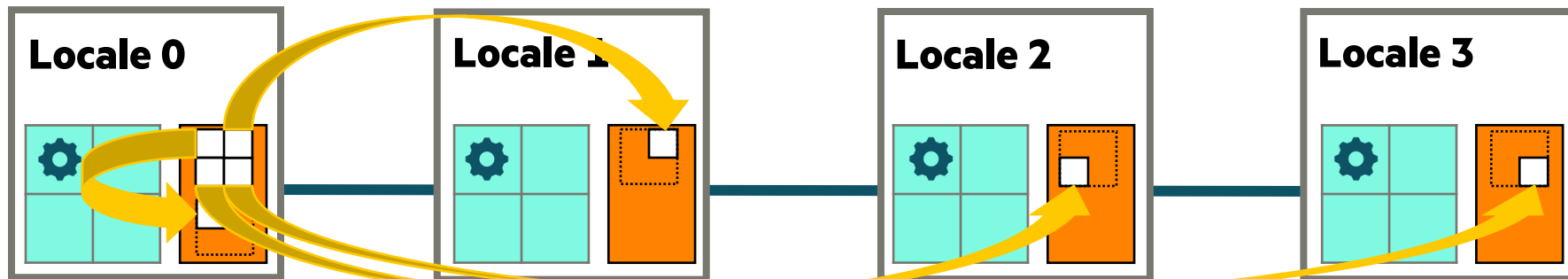
# Array-based Parallelism and Locality

basics-distarr.chpl

```
writeln("Hello from locale ", here.id);  
  
var A: [1..2, 1..2] real;  
  
use BlockDist;  
  
var D = blockDist.createDomain({1..2, 1..2});  
var B: [D] real;  
B = A;
```

Chapel also supports distributed domains (index sets) and arrays

They also result in parallel distributed computation



# Other Chapel Features

- Chapel is a big language
  - everything you'd expect from a modern, productive language
  - plus, additional features supporting parallelism, locality, and scalable performance
- As a result, there are many features you aren't seeing much of today:

## Serial Features:

- **Modules:** for namespacing and code organization
- **Procedures and iterators:** with overloading, generics/polymorphism, rich argument passing, ...
- **OOP:** Value- and Reference-based objects, generic types, inheritance, fields, methods, mix-ins, ...
- ...

## Parallel Features:

- **Rich array support:** multidimensional arrays, sparse arrays, slicing, rank change, reindexing, ...
- **Implicit forms of parallelism:** whole-array operations, promotion of scalar routines, reductions, scans
- **Intra-task synchronization:** atomic and synchronization (full-empty) variables
- ...





# **Sample Compiler Optimizations (Bale IG Revisited)**

# Bale IG in Chapel: Distributed Parallel Version

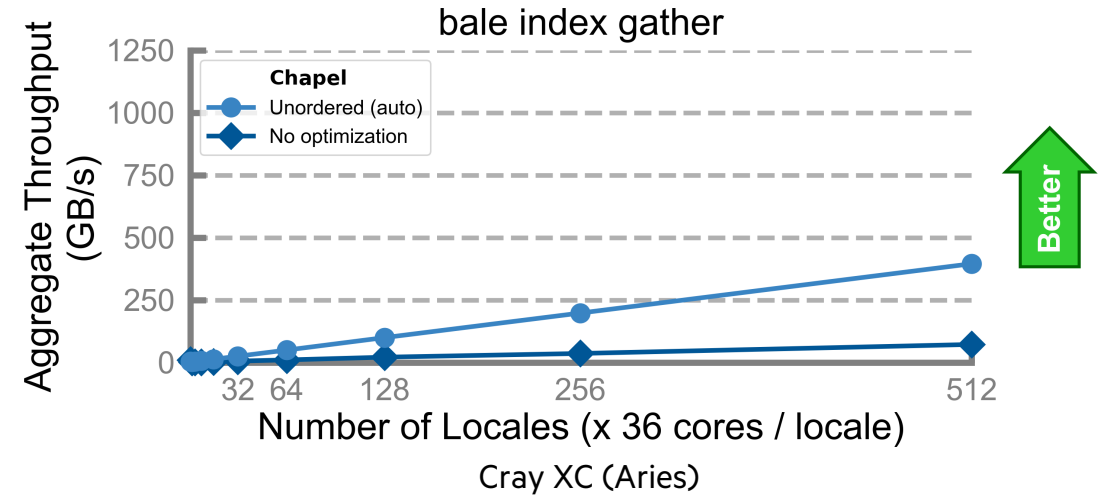
```
use BlockDist;

config const n = 10,
            m = 4;

const SrcInds = blockDist.createDomain(0..<n),
      DstInds = blockDist.createDomain(0..<m);

var Src: [SrcInds] int,
     Inds, Dst: [DstInds] int;
...
forall (d, i) in zip(Dst, Inds) do
    d = Src[i];
```

```
$ chpl bale-ig.chpl
$ ./bale-ig -nl 512
$
```



# Bale IG in Chapel: Distributed Parallel Version (rewrite using parallel iterator)

```
use BlockDist;

config const n = 10,
            m = 4;

const SrcInds = blockDist.createDomain(0..<n),
      DstInds = blockDist.createDomain(0..<m);

var Src: [SrcInds] int,
     Inds, Dst: [DstInds] int;

forall (d, i) in zip(Dst, Inds) do
    d = Src[i];
```

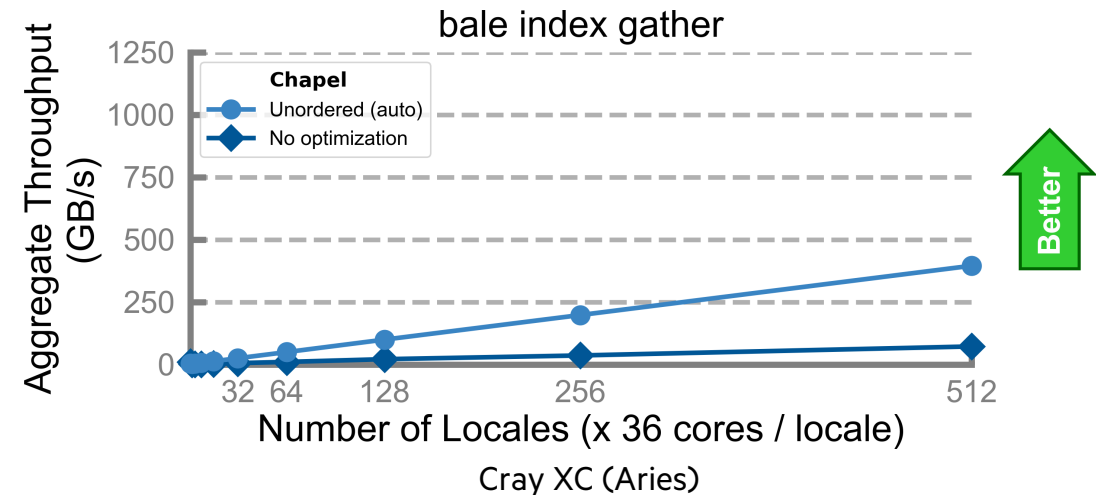
Gets lowered roughly to...

```
$ chp
$ ./b
$
coforall loc in Dst.targetLocales do on loc do
  coforall tid in 0..<here.maxTaskPar do
    foreach idx in myInds(loc, tid, ...) do
      Dst[idx] = Src[Inds[idx]];
```

Create a task per compute node

Create a task per core on that node

Compute that task's gathers



# Bale IG in Chapel: Distributed Parallel Version (optimized using async copies)

```
use BlockDist;

config const n = 10,
            m = 4;

const SrcInds = blockDist.createDomain(0..<n),
      DstInds = blockDist.createDomain(0..<m);

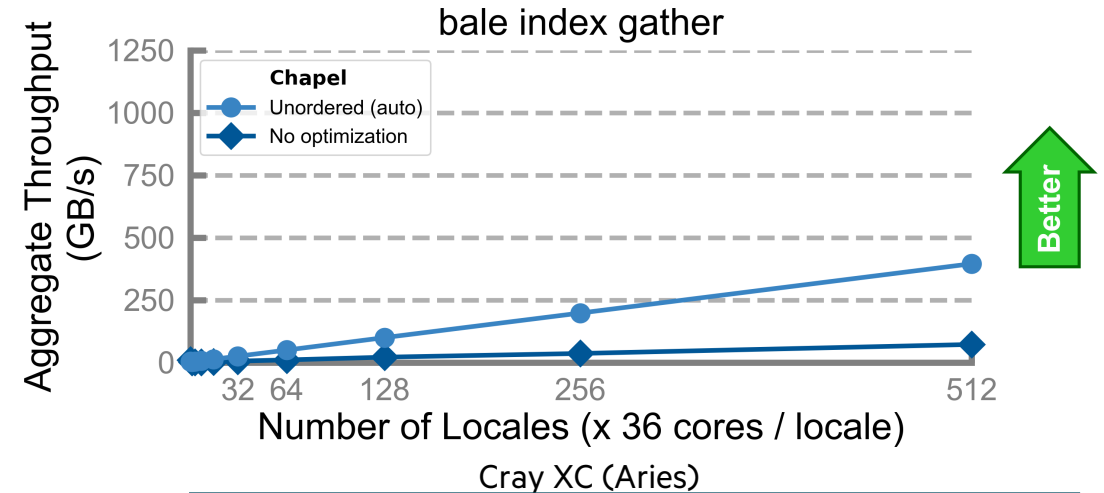
var Src: [SrcInds] int,
     Inds, Dst: [DstInds] int;

forall (d, i) in zip(Dst, Inds) do
    d = Src[i];
```

```
$ chpc
$ ./b
$

coforall loc in Dst.targetLocales do on loc do
    coforall tid in 0..<here.maxTaskPar do
        foreach idx in myInds(loc, tid, ...) do
            Dst[idx] = Src[Inds[idx]];
```

```
foreach idx in myInds(loc, tid, ...) do
    asyncCopy(Dst[idx], Src[Inds[idx]]);
    asyncCopyTaskFence();
```



The user told us this loop was parallel, so why perform these high-latency ops serially?

So, the Chapel compiler rewrites the inner loop to perform them asynchronously

# Bale IG in Chapel: Distributed Parallel Version

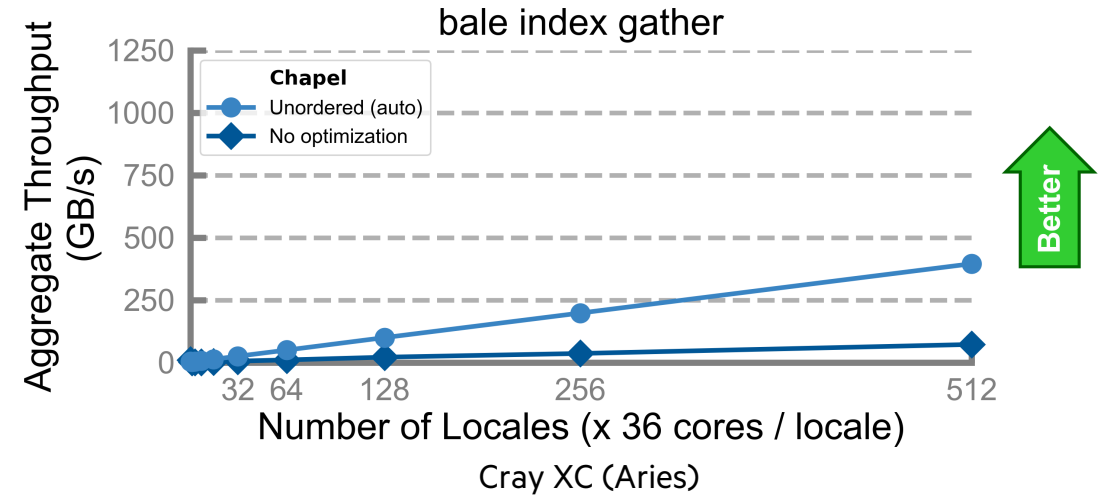
```
use BlockDist;

config const n = 10,
            m = 4;

const SrcInds = blockDist.createDomain(0..<n),
      DstInds = blockDist.createDomain(0..<m);

var Src: [SrcInds] int,
     Inds, Dst: [DstInds] int;
...
forall (d, i) in zip(Dst, Inds) do
    d = Src[i];
```

```
$ chpl bale-ig.chpl
$ ./bale-ig -nl 512
$
```



So far, all communications are being done in a fine-grained manner, an element at a time

# Bale IG in Chapel: Distributed, Explicitly Aggregated Version

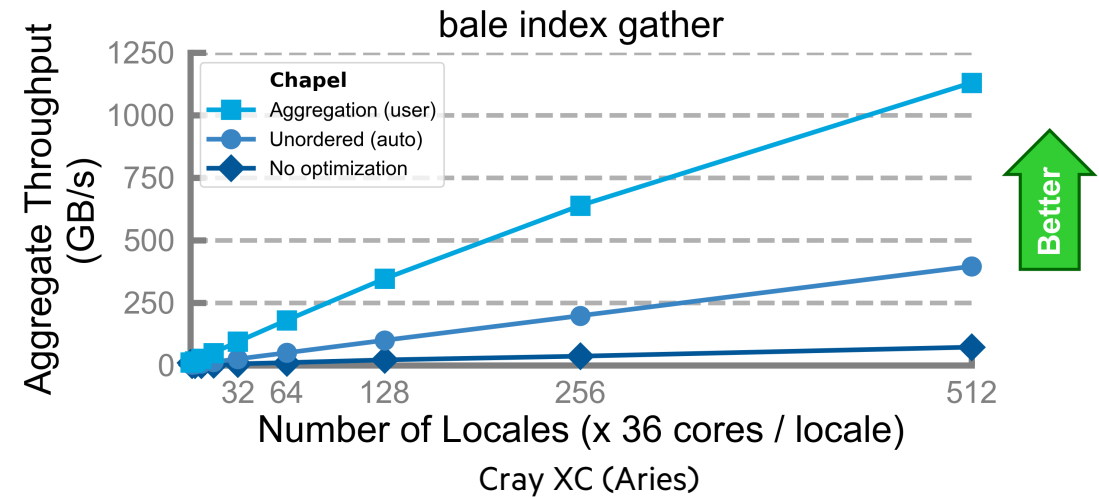
```
use BlockDist, CopyAggregation;

config const n = 10,
             m = 4;

const SrcInds = blockDist.createDomain(0..<n),
       DstInds = blockDist.createDomain(0..<m);

var Src: [SrcInds] int,
     Inds, Dst: [DstInds] int;
...
forall (d, i) in zip(Dst, Inds) with
  (var agg = new SrcAggregator(int)) do
    agg.copy(d, Src[i]);
```

```
$ chpl bale-ig.chpl
$ ./bale-ig -nl 512
$
```



# Bale IG in Chapel: Distributed, Auto-Aggregated Version

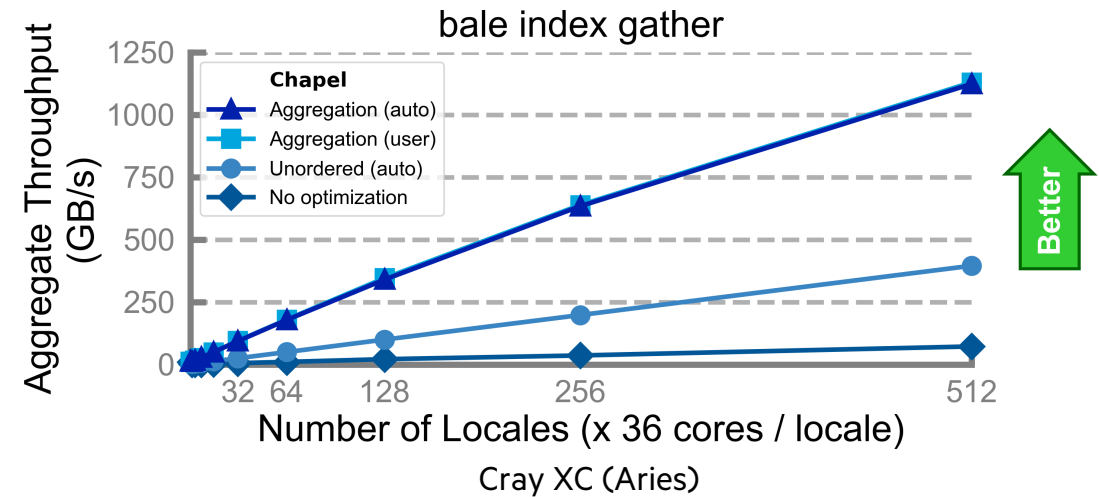
```
use BlockDist;

config const n = 10,
            m = 4;

const SrcInds = blockDist.createDomain(0..<n),
      DstInds = blockDist.createDomain(0..<m);

var Src: [SrcInds] int,
     Inds, Dst: [DstInds] int;
...
forall (d, i) in zip(Dst, Inds) do
    d = Src[i];
```

```
$ chpl bale-ig.chpl --auto-aggregation
$ ./bale-ig -nl 512
$
```



# Bale IG in Chapel vs. SHMEM on Cray XC

## Chapel (Simple / Auto-Aggregated version)

```
forall (d, i) in zip(Dst, Inds) do
    d = Src[i];
```

## Chapel (Explicitly Aggregated version)

```
forall (d, i) in zip(Dst, Inds) with
    (var agg = new SrcAggregator(int)) do
    agg.copy(d, Src[i]);
```

## SHMEM (Exstack version)

```
i=0;
while( exstack_proceed(ex, (i==l_num_req)) ) {
    i0 = i;
    while(i < l_num_req) {
        l_indx = pckindx[i] >> 16;
        pe = pckindx[i] & 0xffff;
        if(!exstack_push(ex, &l_indx, pe))
            break;
        i++;
    }

    exstack_exchange(ex);

    while(exstack_pop(ex, &idx, &fromth)) {
        idx = ltable[idx];
        exstack_push(ex, &idx, fromth);
    }
    lgp_barrier();
    exstack_exchange(ex);

    for(j=i0; j<i; j++) {
        fromth = pckindx[j] & 0xffff;
        exstack_pop_thread(ex, &idx, (uint64_t)fromth);
        tgt[j] = idx;
    }
    lgp_barrier();
}
```

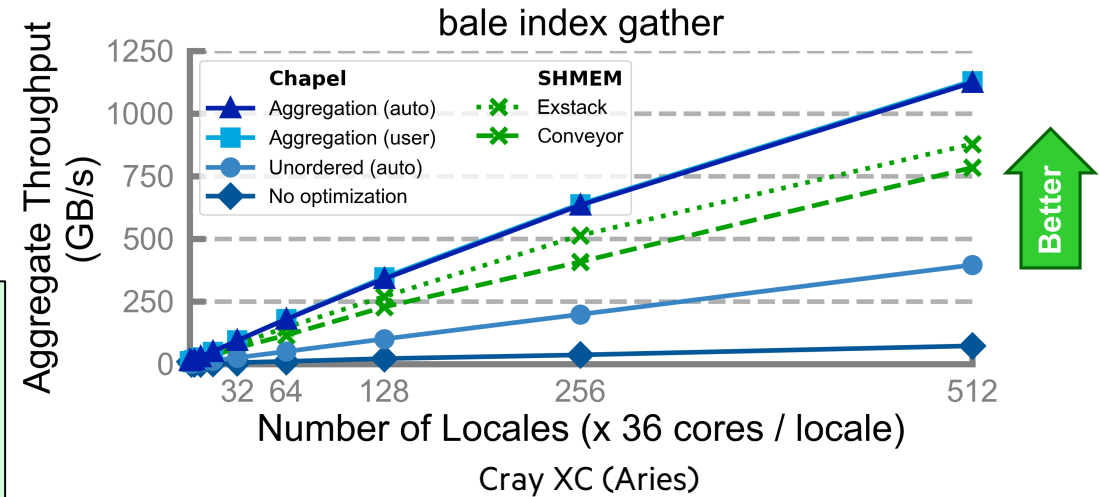
## SHMEM (Conveyors version)

```
i = 0;
while (more = convey_advance(requests, (i == l_num_req)),
        more | convey_advance(replies, !more)) {

    for (; i < l_num_req; i++) {
        pkg.idx = i;
        pkg.val = pckindx[i] >> 16;
        pe = pckindx[i] & 0xffff;
        if (!convey_push(requests, &pkg, pe))
            break;
    }

    while (convey_pull(requests, ptr, &from) == convey_OK) {
        pkg.idx = ptr->idx;
        pkg.val = ltable[ptr->val];
        if (!convey_push(replies, &pkg, from)) {
            convey_unpull(requests);
            break;
        }
    }

    while (convey_pull(replies, ptr, NULL) == convey_OK)
        tgt[ptr->idx] = ptr->val;
}
```





# Bale IG in Chapel vs. SHMEM on HPE Cray EX (Slingshot-11)

## Chapel (Simple / Auto-Aggregated version)

```
forall (d, i) in zip(Dst, Inds) do
    d = Src[i];
```

## Chapel (Explicitly Aggregated version)

```
forall (d, i) in zip(Dst, Inds) with
    (var agg = new SrcAggregator(int)) do
    agg.copy(d, Src[i]);
```

## SHMEM (Exstack version)

```
i=0;
while( exstack_proceed(ex, (i==l_num_req)) ) {
    i0 = i;
    while(i < l_num_req) {
        l_indx = pckindx[i] >> 16;
        pe = pckindx[i] & 0xffff;
        if(!exstack_push(ex, &l_indx, pe))
            break;
        i++;
    }

    exstack_exchange(ex);

    while(exstack_pop(ex, &idx, &fromth)) {
        idx = ltable[idx];
        exstack_push(ex, &idx, fromth);
    }
    lgp_barrier();
    exstack_exchange(ex);

    for(j=i0; j<i; j++) {
        fromth = pckindx[j] & 0xffff;
        exstack_pop_thread(ex, &idx, (uint64_t)fromth);
        tgt[j] = idx;
    }
    lgp_barrier();
}
```

## SHMEM (Conveyors version)

```
i = 0;
while (more = convey_advance(requests, (i == l_num_req)),
        more | convey_advance(replies, !more)) {

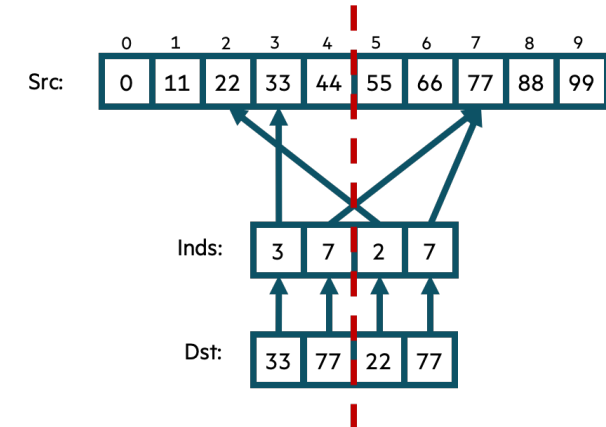
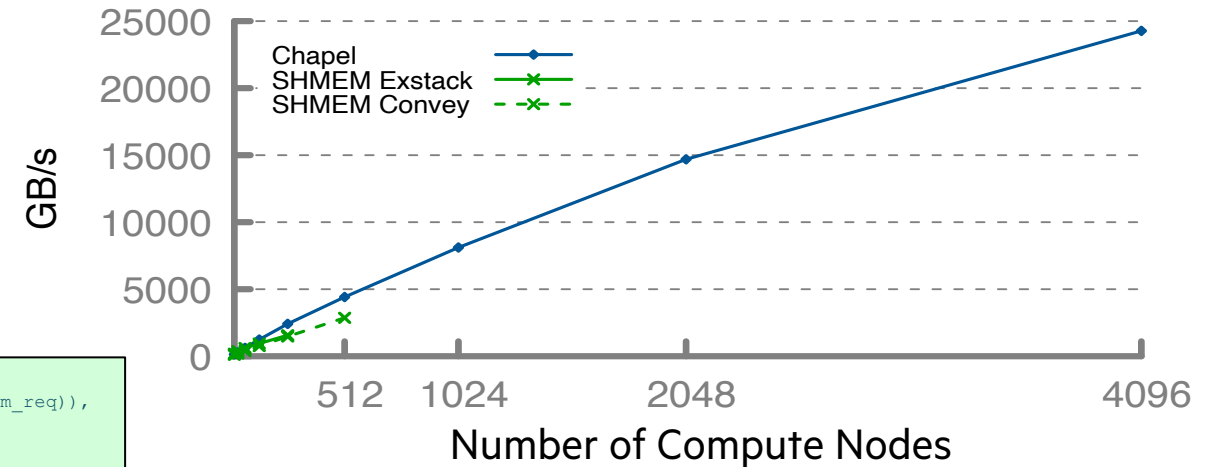
    for (; i < l_num_req; i++) {
        pkg.idx = i;
        pkg.val = pckindx[i] >> 16;
        pe = pckindx[i] & 0xffff;
        if (!convey_push(requests, &pkg, pe))
            break;
    }

    while (convey_pull(requests, ptr, &from) == convey_OK) {
        pkg.idx = ptr->idx;
        pkg.val = ltable[ptr->val];
        if (!convey_push(replies, &pkg, from)) {
            convey_unpull(requests);
            break;
        }
    }

    while (convey_pull(replies, ptr, NULL) == convey_OK)
        tgt[ptr->idx] = ptr->val;
}
```

## Bale Indexgather Performance

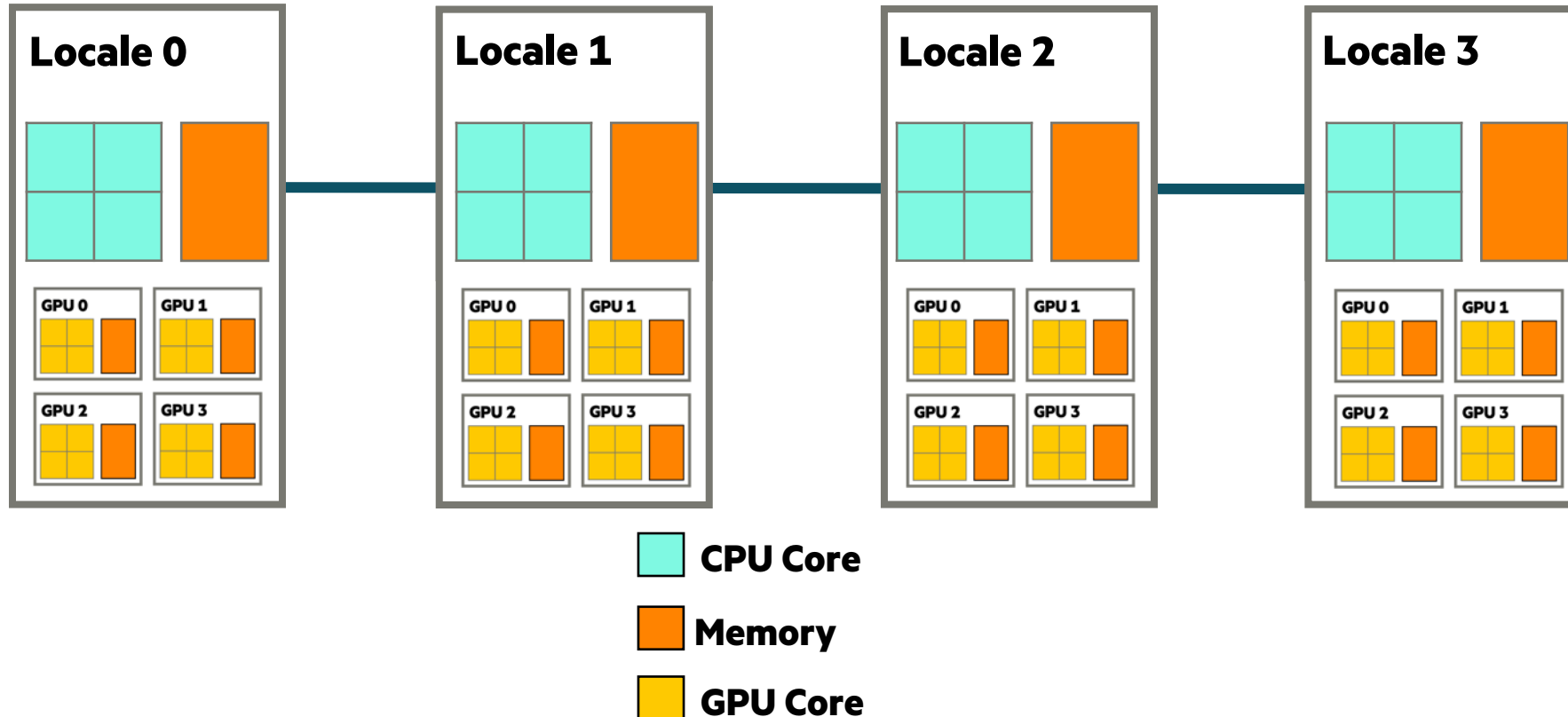
HPE Cray EX (Slingshot-11)



# Programming GPUs with Chapel

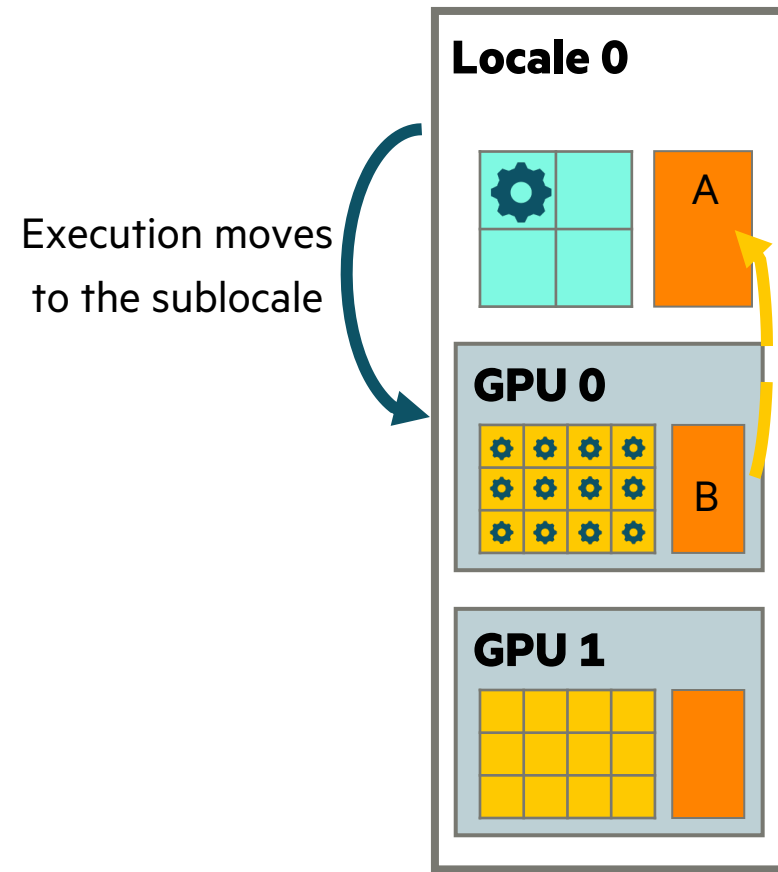
# Representing GPUs in Chapel





- In Chapel, a *locale* refers to a compute resource with processors and memory
  - For now, think of each compute node as being a locale
- Modern systems often involve GPUs as well
  - In Chapel, we represent them as *sub-locals*



# Parallelism and Locality In The Context Of GPUs

 CPU Core    GPU Core    Memory    `var A: [1..n, 1..n] real;`



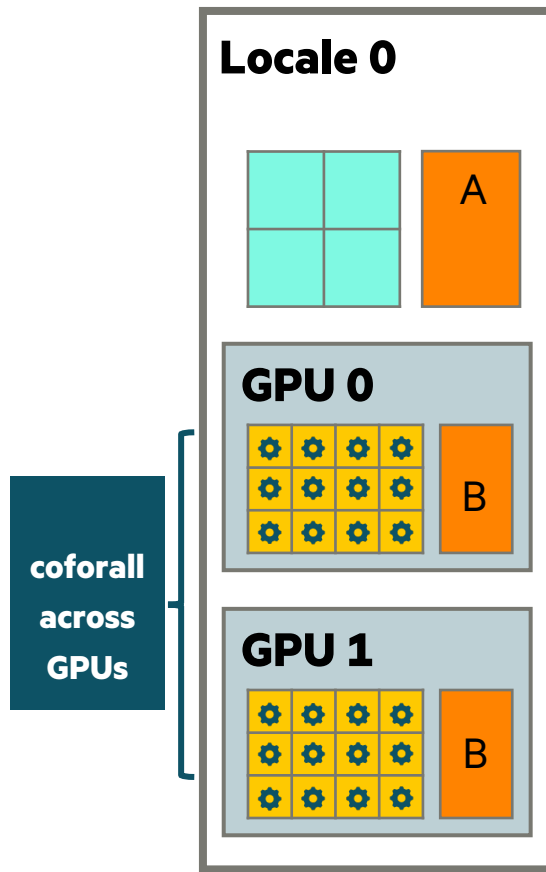
 `on here.gpus[0] {`  
 `var B: [1..n, 1..n] real;`  
 `B = 2;`  
 `A = B;`  
`}`

 `writeln(A);`

# Parallelism and Locality In The Context Of GPUs

 CPU Core    GPU Core    Memory

```
var A: [1..n, 1..n] real;
```

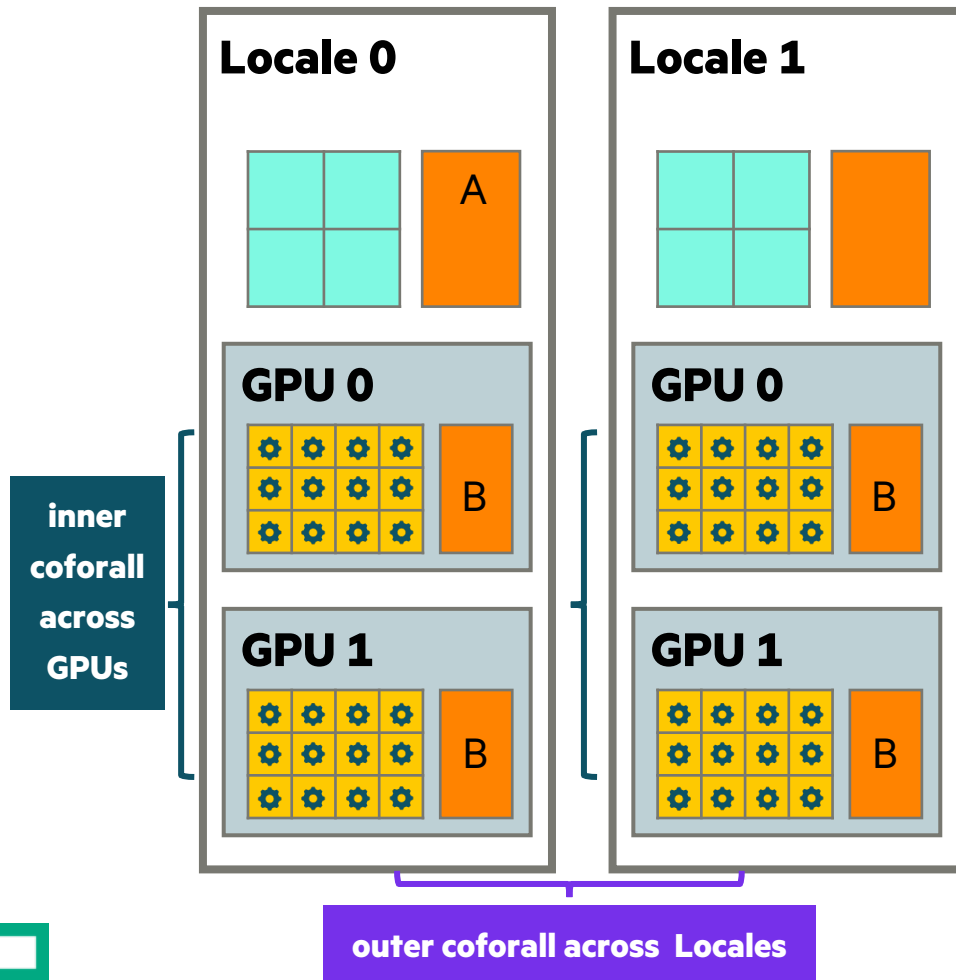


```
coforall g in here.gpus do on g {  
    var B: [1..n, 1..n] real;  
    B = 2;  
    A = B;  
}
```

```
writeln(A);
```

# Parallelism and Locality In The Context Of GPUs

 CPU Core    GPU Core    Memory



```
var A: [1..n, 1..n] real;  
coforall l in Locales do on l {
```

```
coforall g in here.gpus do on g {
```

```
  var B: [1..n, 1..n] real;
```

```
  B = 2;
```

```
  A = B;
```

```
}
```

```
}  
writeln(A);
```

# Parallelism and Locality In The Context Of GPUs

■ CPU Core   
 ■ GPU Core   
 ■ Memory

parallel statements  
with cobegin

Locale 0



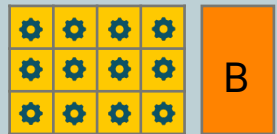
Locale 1



GPU 0



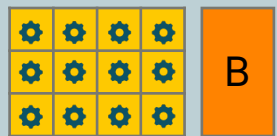
GPU 0



GPU 1



GPU 1



inner  
coforall  
across  
GPUs

outer coforall across Locales

```

var A: [1..n, 1..n] real;
coforall l in Locales do on l {
    cobegin {
        coforall g in here.gpus do on g {
            var B: [1..n, 1..n] real;
            B = 2;
            A = B;
        }
    }
}
writeln(A);
    
```

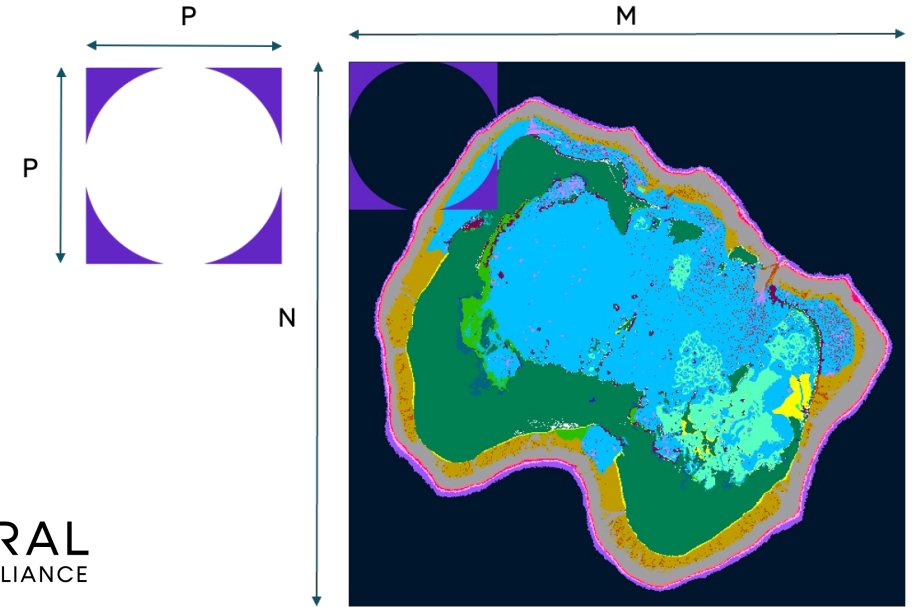
# RapidQ Coral Biodiversity Summary

## What is it?

- Measures coral reef diversity using high-res satellite image analysis
- ~230 lines of Chapel code written in late 2022
- Initial code was CPU-only

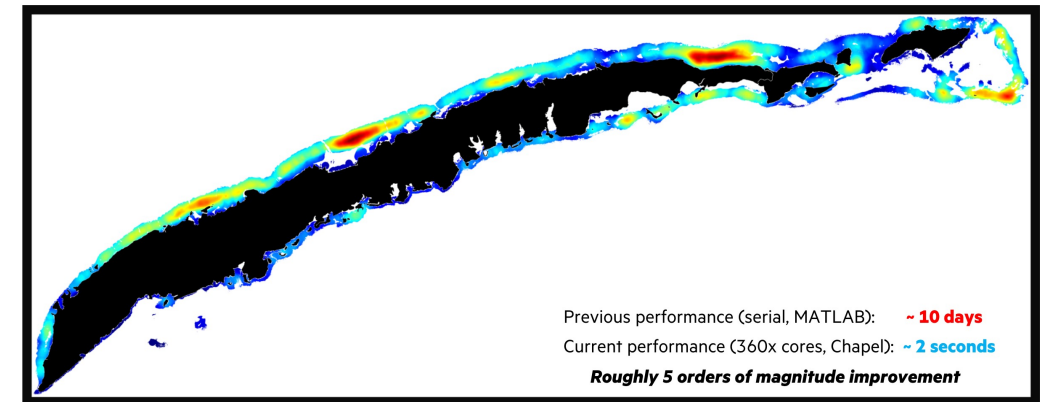
## Who wrote it?

- Scott Bachman, NCAR/[C]Worthy  
– with Rebecca Green, Helen Fox, Coral Reef Alliance



## Why Chapel?

- easy transition from Python, which was being used
- massive performance improvement:  
~10-day Python run finished in ~2 seconds using 360 cores
- enabled unexpected algorithmic improvements

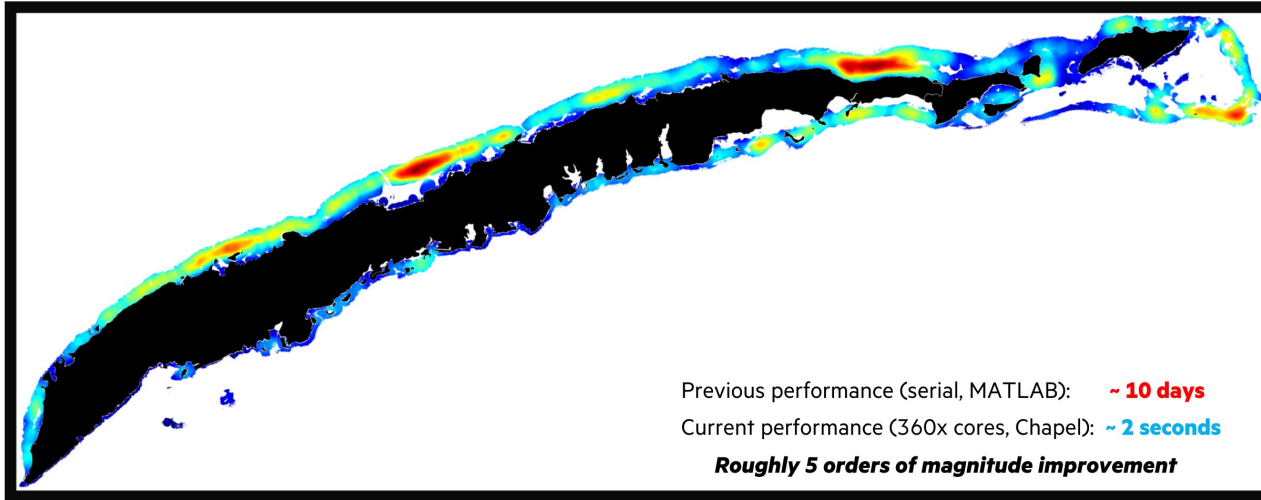


From Scott Bachman's CHI UW 2023 talk: <https://youtu.be/IJhh9KLL2X0>



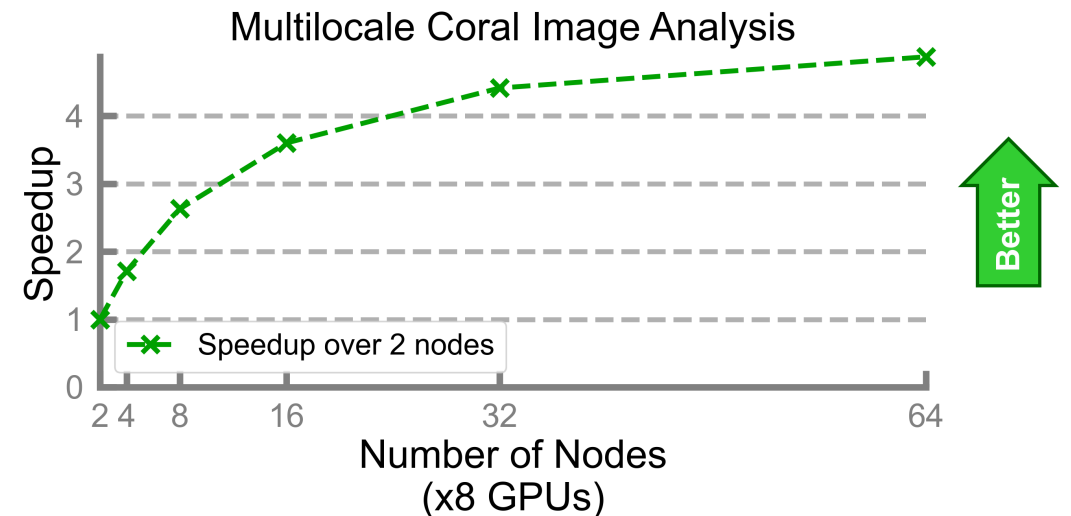
# Coral Reef Spectral Biodiversity: Productivity and Performance

**Original algorithm:** Habitat Diversity,  $O(M \cdot N \cdot P)$



**Improved algorithm:** Spectral Diversity,  $O(M \cdot N \cdot P^3)$

- Chapel run was estimated to require ~4 weeks on 8-core desktop
- updated code to leverage GPUs
  - required adding ~90 lines of code for a total of ~320
- ran in ~20 minutes on 64 nodes of Frontier
  - 512 NVIDIA K20X Kepler GPUs



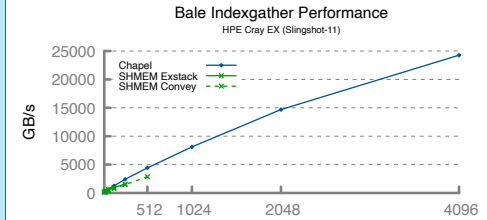
## Wrap-up

# Chapel Summary

## Chapel is unique among programming languages

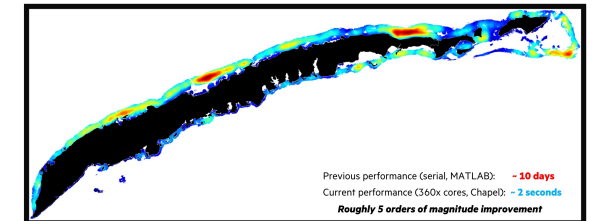
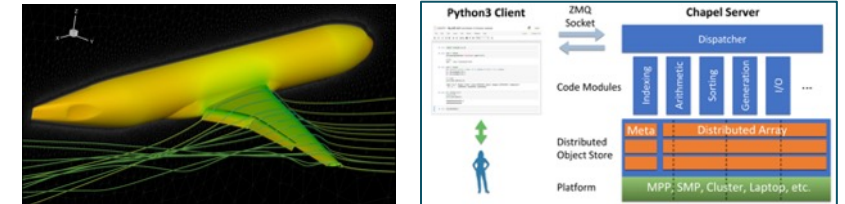
- built-in features for scalable parallel computing make it HPC-ready
- supports clean, concise code relative to conventional approaches
- ports and scales from laptops to supercomputers
- supports GPUs in a vendor-neutral manner

```
use BlockDist;  
  
config const n = 10,  
           m = 4;  
  
const SrcInds = blockDist.createDomain(0..  
    n),  
       DstInds = blockDist.createDomain(0..  
    m);  
  
var Src: [SrcInds] int,  
     Inds, Dst: [DstInds] int;  
  
...  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```



## Chapel is being used for productive parallel computing at scale

- users are reaping its benefits in practical, cutting-edge applications
- applicable to domains as diverse as physical simulations and data science
- Arkouda is a particularly unique example of driving HPCs from Python



# Takeaways for this PLSE setting

For scalable parallel computing, good language design can...

...**provide built-in abstractions** to simplify the expression of parallel operations

– e.g., parallel loops and iterators, global namespace

...**more clearly represent parallel computations** compared to standard approaches

– e.g., MPI, SHMEM, CUDA, HIP, SYCL, OpenMP, OpenCL, OpenACC, Kokkos, RAJA, ...

...permit users to **create new abstractions** supporting performance and/or clean code

– e.g., per-task aggregators

...**enable new optimization opportunities** by expressing parallelism and locality clearly

– e.g., asynchronous operations, auto-aggregation of communication

...**support excellent performance and scalability**

– e.g., to thousands of nodes and over a million cores



# The Chapel Team at HPE









# Ways to Engage with the Chapel Community

## Live/Virtual Events

- [ChapelCon](#) (formerly CHI UW), annually
- [Office Hours](#), monthly
- [Live Demo Sessions](#), monthly



## Community / User Forums

- [Discord](#)  **Discord**
- [Discourse](#)  **Discourse**  
chapel+qs@discoursemail.com
- Email Contact Alias
- [GitHub Issues](#) 
- [Gitter](#)  **GITTER**
- [Reddit](#)  **reddit**
- [Stack Overflow](#)  **stackoverflow**

## Electronic Broadcasts

- [Chapel Blog](#), ~biweekly
- [Community Newsletter](#), quarterly
- [Announcement Emails](#), around big events

## Social Media

- [Bluesky](#) 
- [Facebook](#) 
- [LinkedIn](#) 
- [Mastodon](#) 
- [X / Twitter](#) 
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# Thank you

---

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