



**Hewlett Packard
Enterprise**

PARALLEL PROGRAMMING IN CHAPEL: OVERVIEW AND OOKAMI



Brad Chamberlain

Webinar for the Ookami user community

January 13, 2022



TEASER FOR THIS TALK

Imagine having a programming language for HPC that was as...

...**programmable** as Python

...yet also as...

...**fast** as Fortran

...**scalable** as MPI or SHMEM

...**portable** as C

...**flexible** as C++

...**type-safe** as Fortran, C, C++, ...

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



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



WHAT DOES “PRODUCTIVITY” MEAN TO YOU?

Recent Graduates:

“Something similar to what I used in school: Python, Matlab, Java, ...”

Seasoned HPC Programmers:

“That sugary stuff which I can’t use because I need full control to ensure good performance”

Computational Scientists:

“Something that lets me focus on my science without having to wrestle with architecture-specific details”

Chapel Team:

“Something that lets computational scientists express what they want, without taking away the control that HPC programmers need, implemented in a language that’s attractive to recent graduates.”



SPEAKING OF THE CHAPEL TEAM...

Chapel is truly a team effort—we're currently at 19 full-time employees (+ a director), and we are hiring

Chapel Development Team at HPE



see: <https://chapel-lang.org/contributors.html>



OUTLINE

An aerial photograph of a rugged coastline. In the center, a rectangular swimming pool is built into a rocky cliffside. The pool's water is a vibrant greenish-yellow. To the right of the pool, a sandy beach is visible, with several people walking or standing. The ocean is a deep blue, with white waves crashing against the dark, moss-covered rocks. The overall scene is a mix of natural beauty and human-made structures.

I. What is Chapel?

II. Chapel Motivation and Uses

III. Introduction to Chapel

IV. Live Demo?

V. Chapel on Ookami

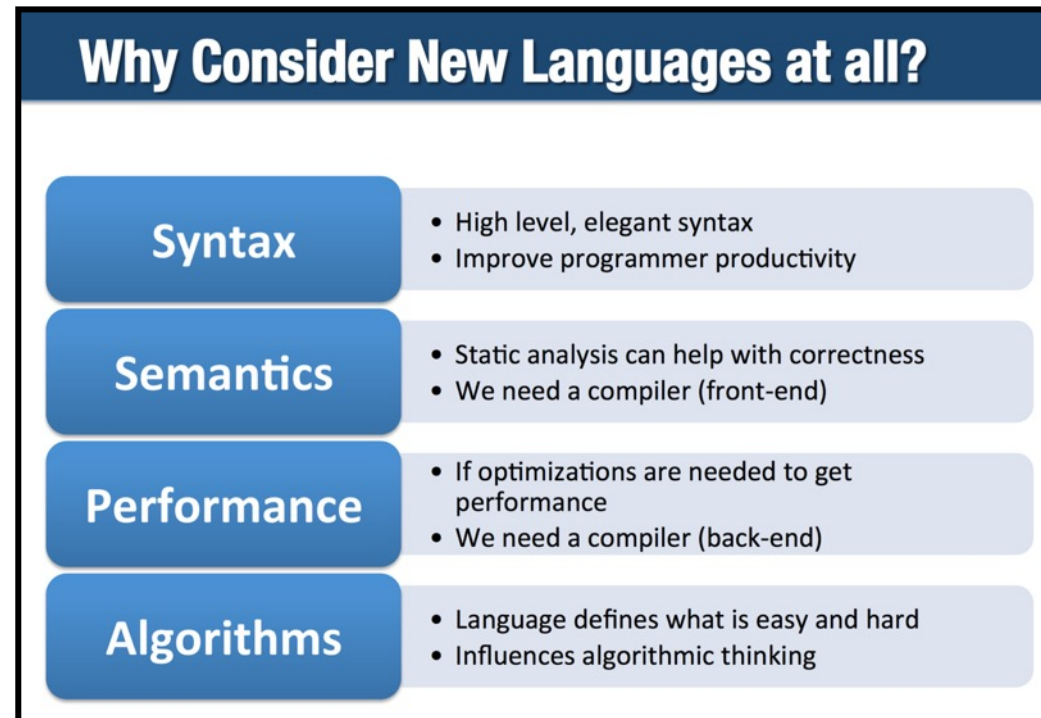
V. Summary and Resources

VI. Hands-On?

WHY CREATE A NEW LANGUAGE?

- **Because parallel programmers deserve better**

- the state of the art for HPC is a mish-mash of libraries, pragmas, and extensions
- parallelism and locality are concerns that deserve first-class language features



[Image Source:
Kathy Yelick's (UC Berkeley, LBNL)
[CHIUV 2018](#) keynote:
[Why Languages Matter More Than Ever](#),
used with permission]

- **And because existing languages don't meet our needs...**



WHAT SHOULD A PRODUCTIVE LANGUAGE FOR HPC SUPPORT?

Traditional Language Characteristics

- programmability
- portability
- performance
- abstraction
- interoperability
- ...

Features Unique to HPC

- ability to express **parallelism**
- ability to control and reason about **locality**

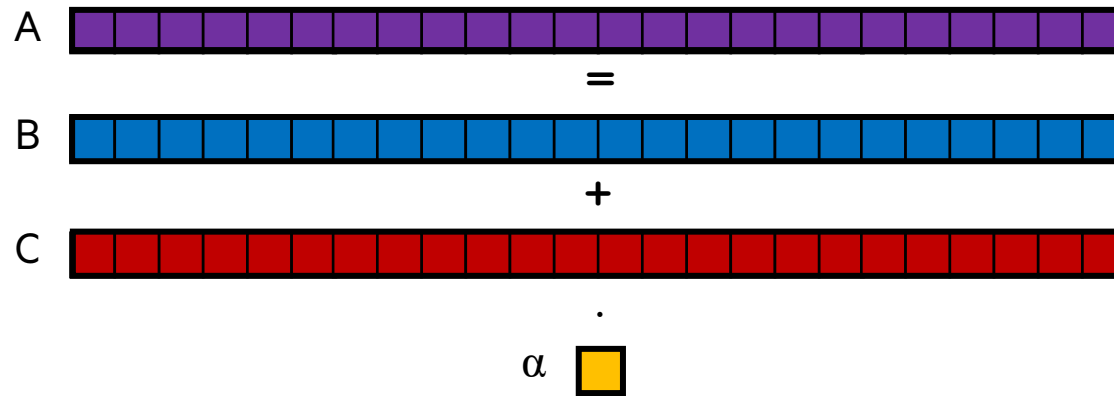


STREAM TRIAD: A TRIVIAL CASE OF PARALLELISM + LOCALITY

Given: m -element vectors A, B, C

Compute: $\forall i \in 1..m, A_i = B_i + \alpha \cdot C_i$

In pictures:

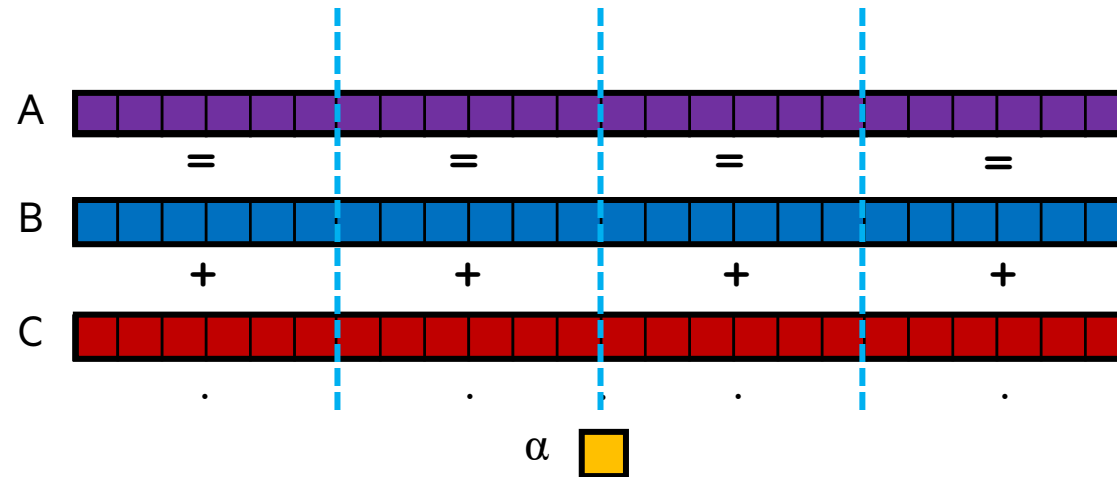


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In pictures, in parallel (shared memory / multicore):

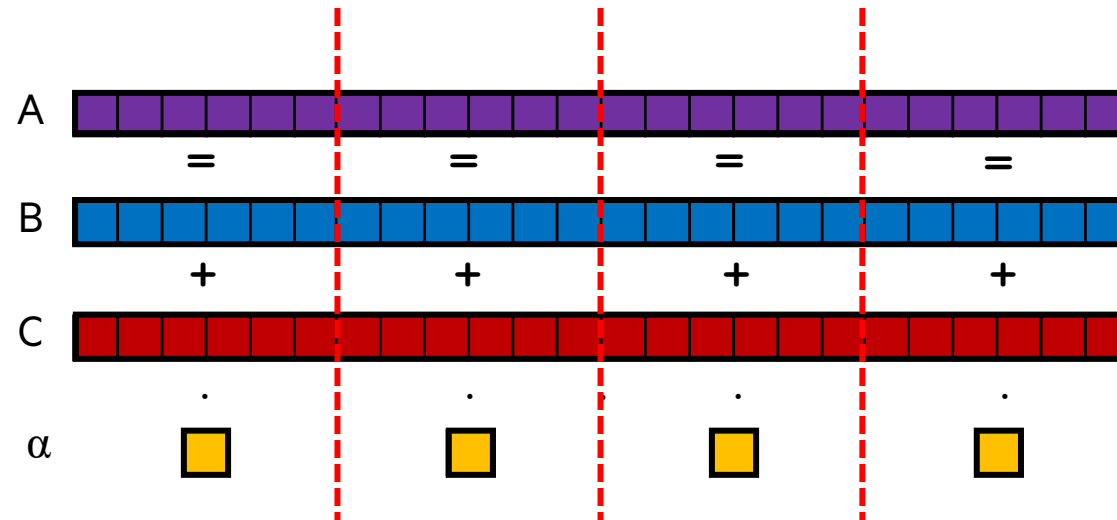


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Given: m -element vectors A, B, C

Compute: $\forall i \in 1..m, A_i = B_i + \alpha \cdot C_i$

In pictures, in parallel (distributed memory):

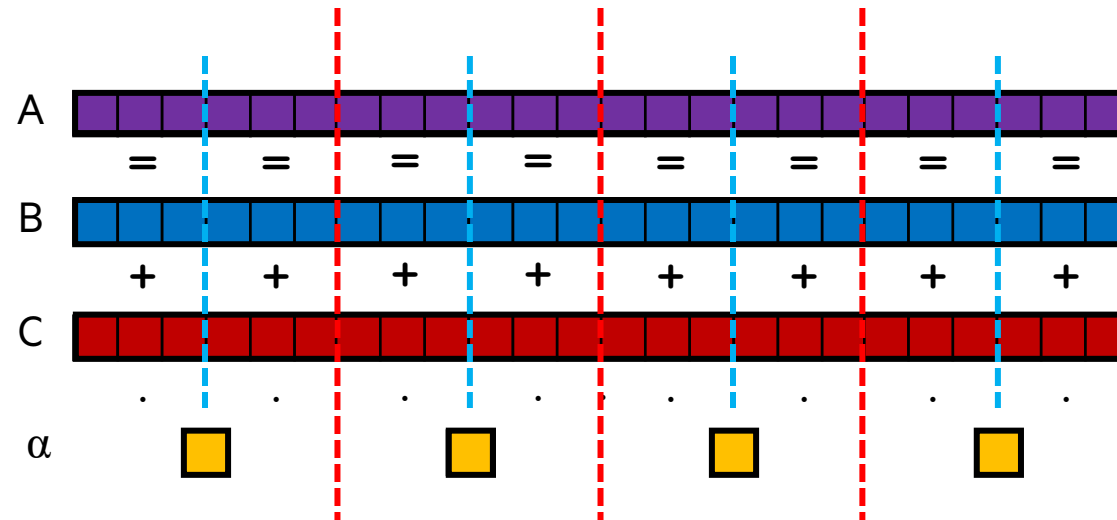


STREAM TRIAD: A TRIVIAL CASE OF PARALLELISM + LOCALITY

Given: m -element vectors A, B, C

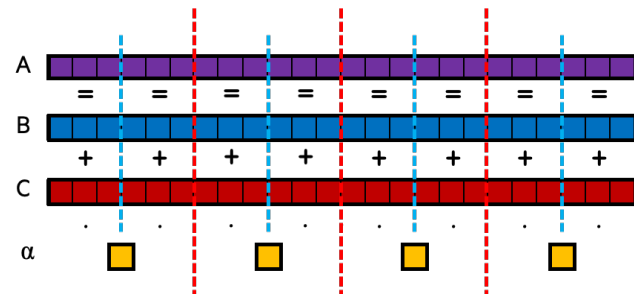
Compute: $\forall i \in 1..m, A_i = B_i + \alpha \cdot C_i$

In pictures, in parallel (distributed memory multicore):



STREAM TRIAD IN CONVENTIONAL HPC PROGRAMMING MODELS

Many Disparate Notations for Expressing Parallelism + Locality



```
#include <hpcc.h> MPI

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 );
        }
    }

    for (j=0; j<VectorSize; j++) {
        b[j] = 2.0;
        c[j] = 1.0;
    }
    scalar = 3.0;

    for (j=0; j<VectorSize; j++)
        a[j] = b[j]+scalar*c[j];

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

    return 0; }

```

Note: This is a very trivial parallel computation—imagine the additional differences for something more complex!

Challenge: Can we do better?



BALE INDEX GATHER: CHAPEL VS. EXSTACK VS. CONVEYORS

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();
}

```

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;
}

```

Elegant Chapel version (compiler-optimized w/ '--auto-aggregation')

```

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

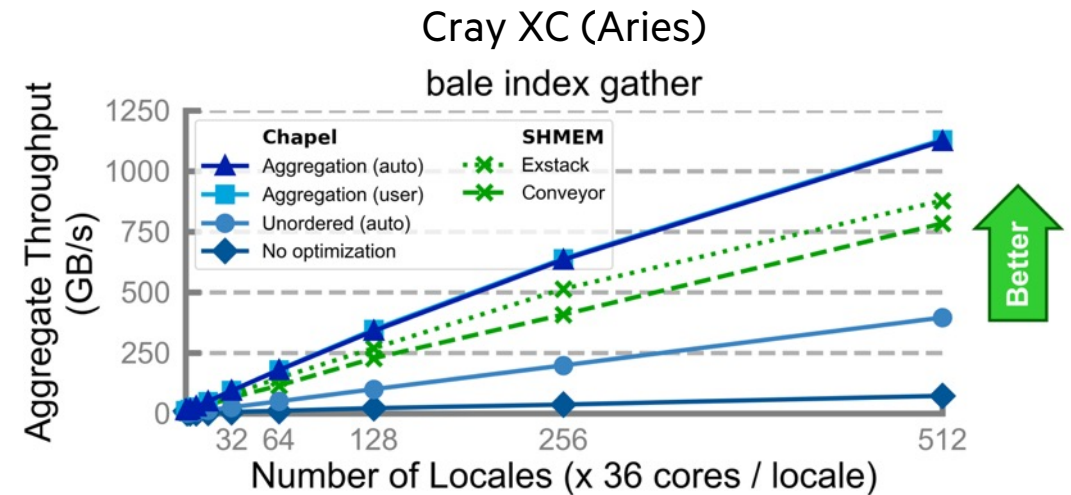
```

Manually Tuned Chapel version (using aggregator abstraction)

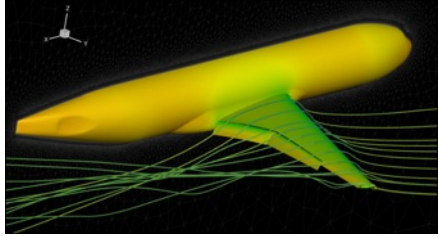
```

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

```

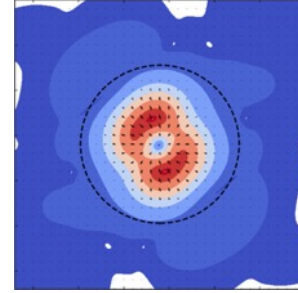


CURRENT FLAGSHIP CHAPEL APPLICATIONS



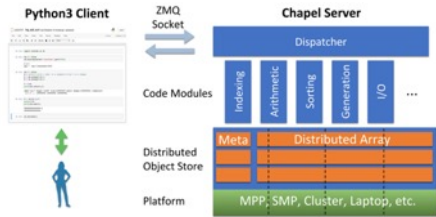
CHAMPS: 3D Unstructured CFD

Éric Laurendeau, Simon Bourgault-Côté,
Matthieu Parenteau, et al.
École Polytechnique Montréal



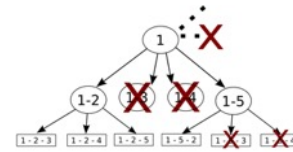
ChpUltra: Simulating Ultralight Dark Matter

Nikhil Padmanabhan, J. Luna Zagorac, et al.
Yale University / University of Auckland



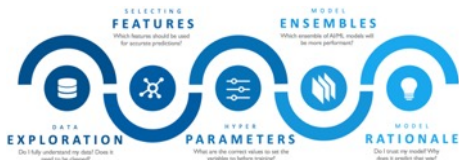
Arkouda: NumPy at Massive Scale

Mike Merrill, Bill Reus, et al.
US DoD



ChOp: Chapel-based Optimization

Tiago Carneiro, Nouredine Melab, et al.
INRIA Lille, France



CrayAI: Distributed Machine Learning

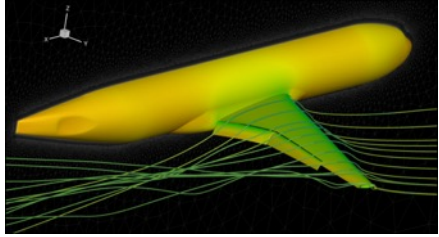
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Your application here?

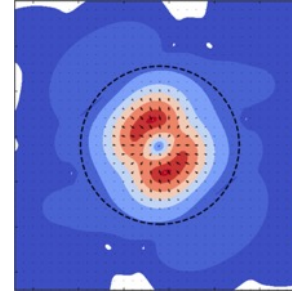


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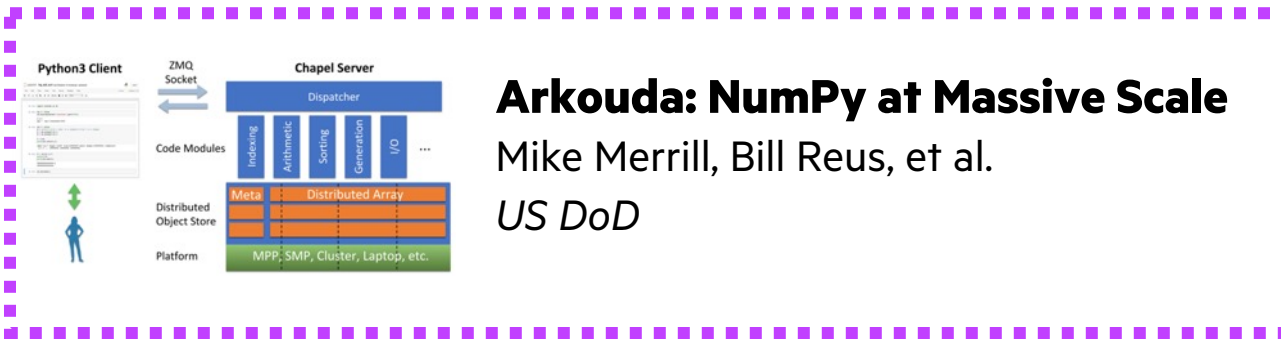
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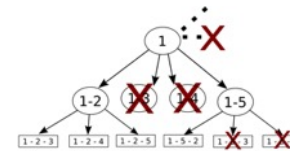
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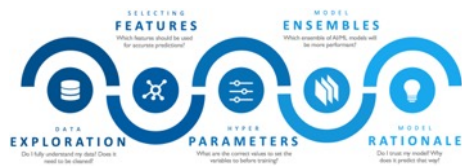
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Hewlett Packard Enterprise



Your application here?



ARKOUDA IN ONE SLIDE

What is it?

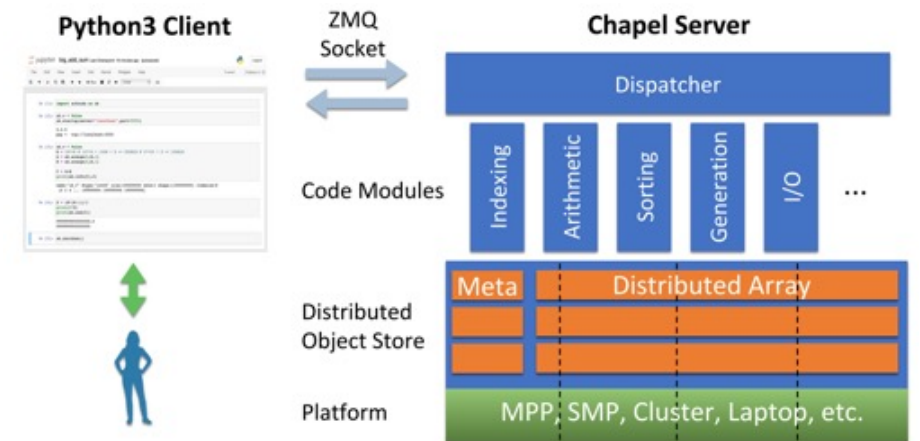
- A Python library supporting a key subset of NumPy and Pandas for Data Science
 - Computes massive-scale results within the human thought loop (seconds to minutes on multi-TB-scale arrays)
 - Uses a Python-client/Chapel-server model to get scalability and performance
- ~16k lines of Chapel, largely written in 2019, continually improved since then

Who wrote it?

- Mike Merrill, Bill Reus, et al., US DoD
- Open-source: <https://github.com/Bears-R-Us/arkouda>

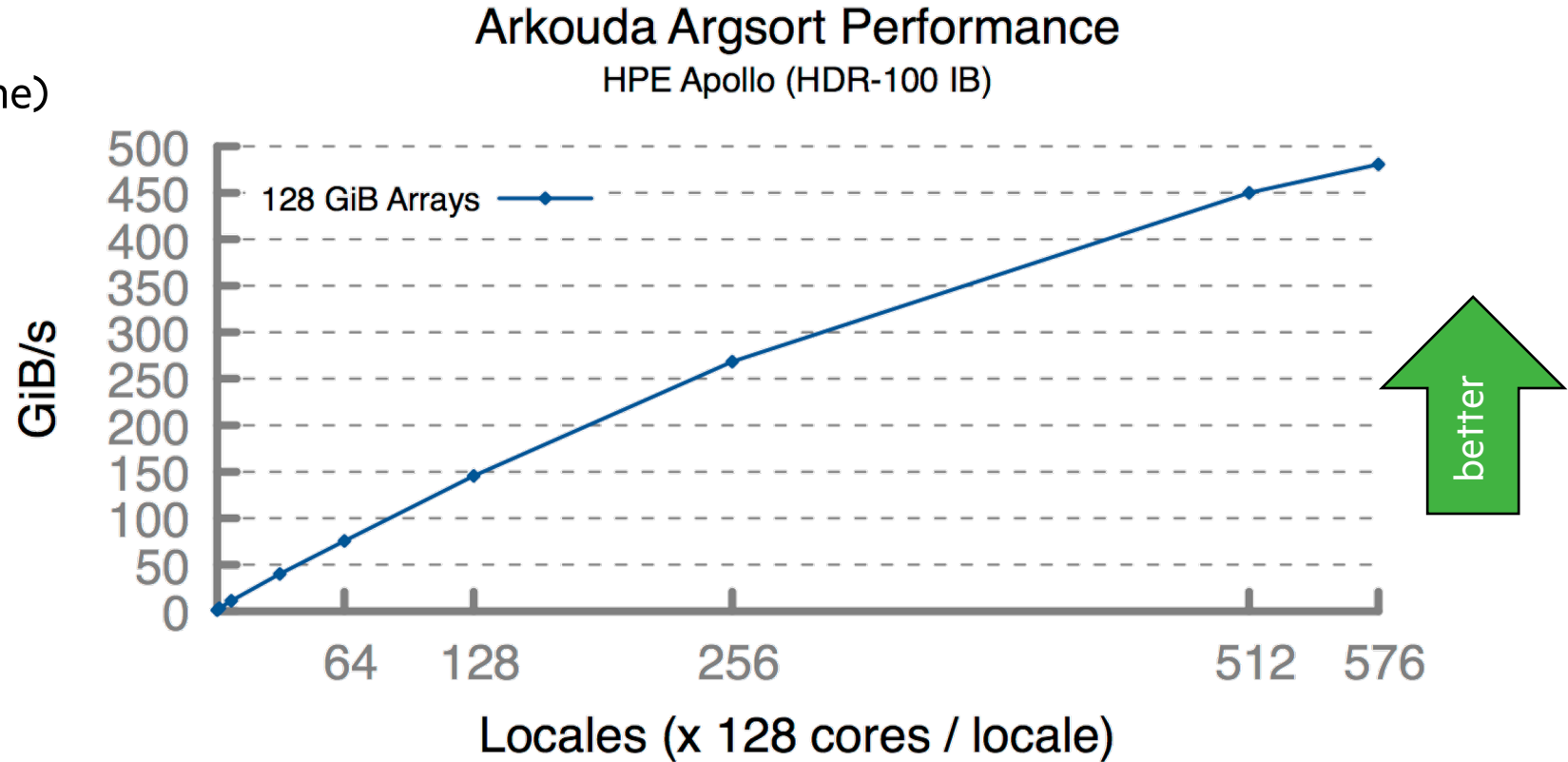
Why Chapel?

- high-level language with performance and scalability
 - close to Pythonic—doesn't repel Python users who look under the hood
- great distributed array support
- ports from laptop to supercomputer



ARKOUDA ARGSORT: HERO RUN

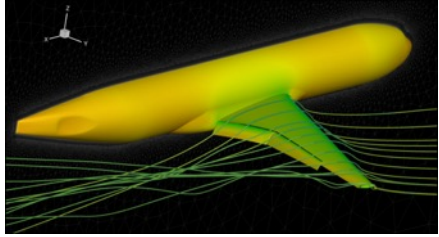
- Recent run performed on a large Apollo system
 - 72 TiB of 8-byte values
 - 480 GiB/s (2.5 minutes elapsed time)
 - used 73,728 cores of AMD Rome
 - ~100 lines of Chapel code



Close to world-record performance—Quite likely a record for performance::lines of code

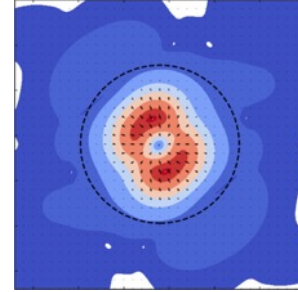


CURRENT FLAGSHIP CHAPEL APPLICATIONS



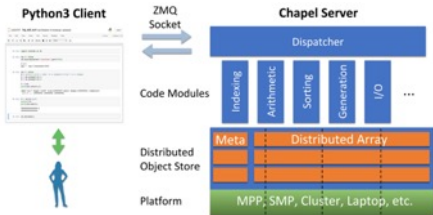
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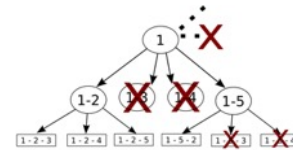
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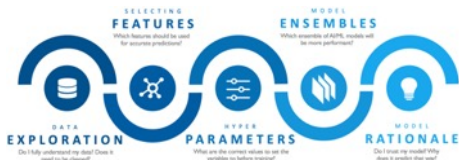
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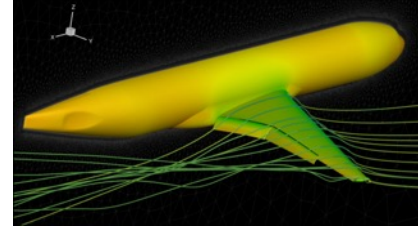
Your application here?



CHAMPS SUMMARY

What is it?

- 3D unstructured CFD framework for airplane simulation
- ~48k lines of Chapel written from scratch in ~2 years



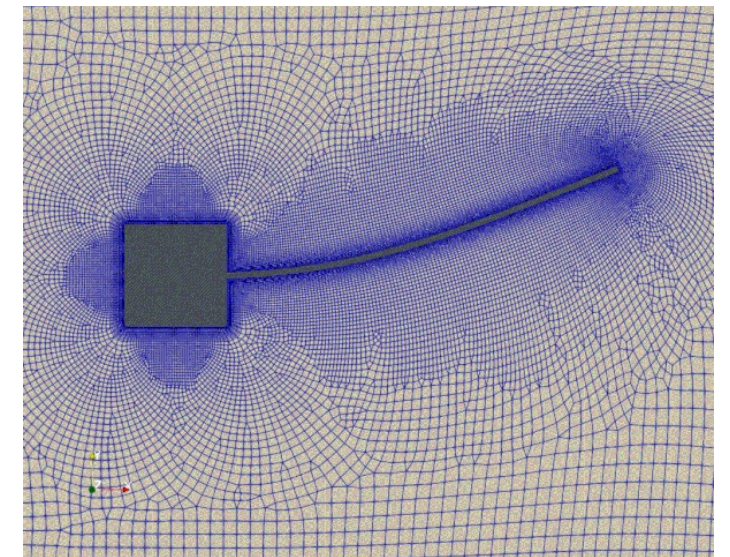
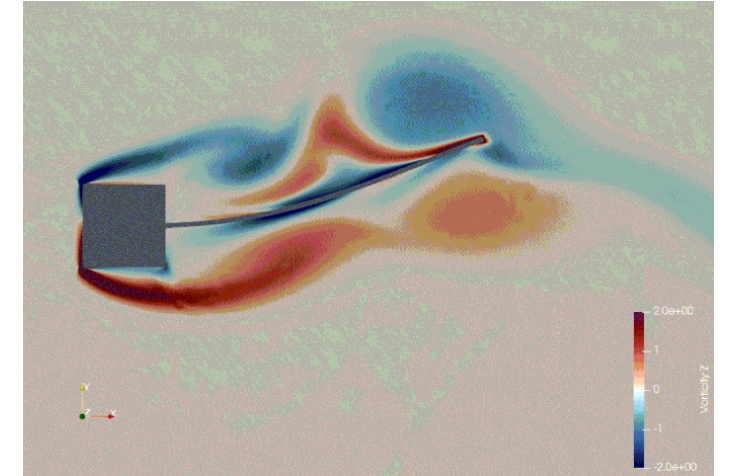
Who wrote it?

- Professor Éric Laurendeau's team at Polytechnique Montreal



Why Chapel?

- performance and scalability competitive with MPI + C++
- students found it far more productive to use



CHAMPS: EXCERPT FROM ERIC'S CHIUW 2021 KEYNOTE (VIDEO)

HPC Lessons From 30 Years of Practice in CFD Towards Aircraft Design and Analysis

LAB HISTORY AT POLYTECHNIQUE

- **NSCODE** (2012 - early 2020):
 - Shared memory 2D/2.5D structured multi-physics solver written in C/Python
 - ~800 C/header files: ~120k lines of code
 - Run by Python interface using f2py (f90 APIs)
 - Difficult to maintain at the end or even to merge new developments
- **(U)VLM** (2012 - now):
 - ~5-6 versions in different languages (Matlab, Fortran, C++, Python, Chapel)
 - The latest version in Chapel is integrated in CHAMPS
- **EULER2D** (early 2019):
 - Copy in Chapel of a small version of NSCODE as benchmark between C and Chapel that illustrated the Chapel language potential
 - ~10 Chapel files: ~1750 lines of code
- **CHAMPS** (mid 2019 - now):
 - Distributed memory 3D/2D unstructured multi-physics solver written in Chapel
 - ~120 Chapel files: ~48k lines of code



25



**POLYTECHNIQUE
MONTRÉAL**

https://youtu.be/wD-a_KyB8aI?t=1904

CHAMPS: EXCERPT FROM ERIC'S CHIUW 2021 KEYNOTE

HPC Lessons From 30 Years of Practice in CFD Towards Aircraft Design and Analysis

“To show you what Chapel did in our lab... [NSCODE, our previous framework] ended up 120k lines. And my students said, ‘We can't handle it anymore. It's too complex, we lost track of everything.’ And today, they went **from 120k lines to 48k lines, so 3x less.**

But the code is not 2D, it's 3D. And it's not structured, it's unstructured, which is way more complex. And it's multi-physics: aeroelastic, aero-icing. **So, I've got industrial-type code in 48k lines.**

So, for me, this is like the proof of the benefit of Chapel, **plus the smiles I have on my students everyday in the lab because they love Chapel as well.** So that's the key, that's the takeaway.

[Chapel] promotes the programming efficiency ... **We ask students at the master's degree to do stuff that would take 2 years and they do it in 3 months.** So, if you want to take a summer internship and you say, ‘program a new turbulence model,’ well they manage. And before, it was impossible to do.”

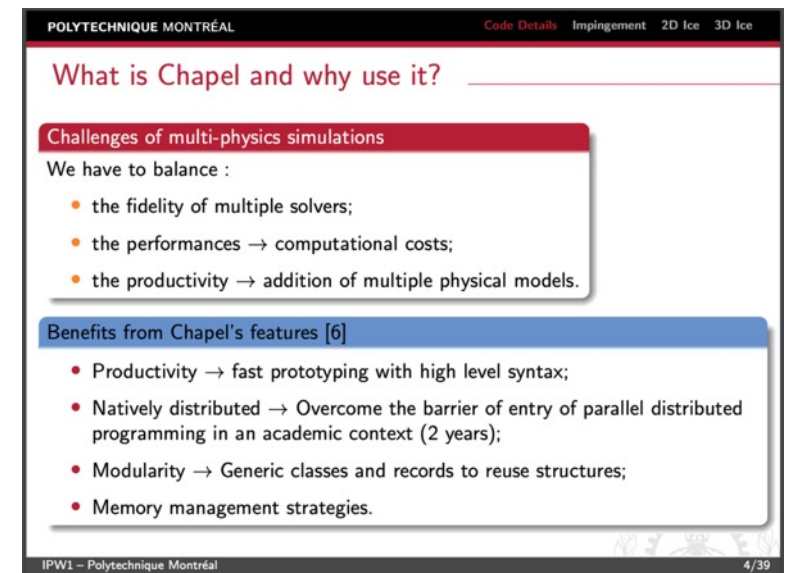
- Talk available online: https://youtu.be/wD-a_KyB8aI?t=1904 (hyperlink jumps to the section quoted here)



**POLYTECHNIQUE
MONTRÉAL**

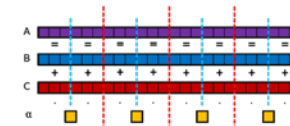
CHAMPS 2021 HIGHLIGHTS

- Presented at CASI/IASC Aero 21 Conference
 - Participated in 1st AIAA Ice Prediction Workshop
 - Participating in 4th AIAA CFD High-lift Prediction Workshop
 - Student presentation to CFD Society of Canada (CFDSC)
-
- **Achieving large-scale, high-quality results comparable to other major players in industry, government, academia:**
 - e.g., Boeing, Lockheed Martin, NASA, JAXA, Georgia Tech, ...



SUMMARY OF THIS SECTION

- Conventional HPC programming notations are not particularly productive
 - they utilize too many distinct ways of specifying locality and parallelism
 - they are too specific to certain flavors of locality or parallelism

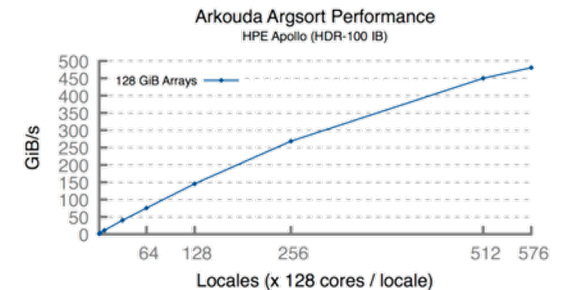
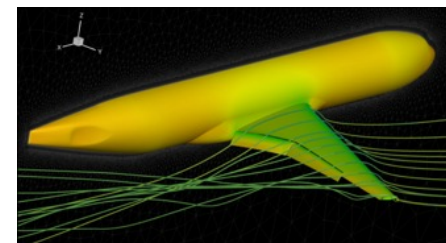
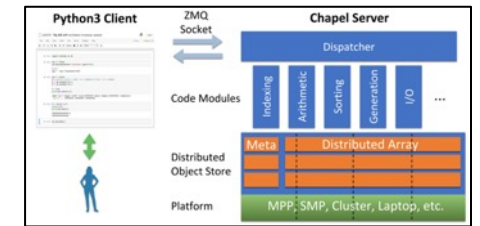
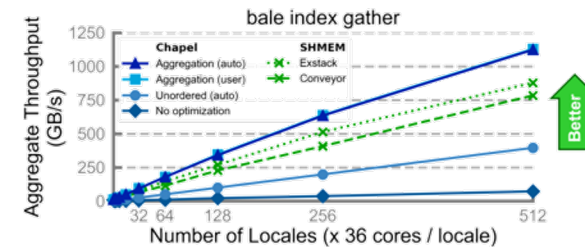


```

MPI + OpenMP
#include <hpc.h>
#ifdef _OPENMP
#include <omp.h>
#endif
static int VectorSize;
static double *a, *b, *c;
int rv, errCount;
int HPC_Stream(HPC_Params *params) {
    int myRank, commSize;
    MPI_Comm comm = MPI_COMM_WORLD;
    MPI_Comm_size(comm, commSize);
    MPI_Comm_rank(comm, myRank);
    rv = HPC_Stream(params, 0 == myRank);
    MPI_Reduce(&rv, &errCount, 1, MPI_INT, MPI_SUM,
              0, comm);
    return errCount;
}
int HPC_Stream(HPC_Params *params, int doIt) {
    register int i;
    double scalar;
    VectorSize = HPC_LocalVectorSize(params, 3,
    sizeof(double) * 3);
    a = HPC_MALLOC(double, VectorSize);
    b = HPC_MALLOC(double, VectorSize);
    c = HPC_MALLOC(double, VectorSize);
}

CUDA
#define N 200000
int main() {
    float *d_a, *d_b, *d_c;
    float scalar;
    cudaMalloc((void**) &d_a, sizeof(float)*N);
    cudaMalloc((void**) &d_b, sizeof(float)*N);
    cudaMalloc((void**) &d_c, sizeof(float)*N);
    dim3 dimBlock(128);
    dim3 dimGrid(N/dimBlock.x);
    if (N % dimBlock.x != 0) dimGrid.x--;
    set_array<cudaGrid, dimBlock>>(d_b, .5f, N);
    set_array<cudaGrid, dimBlock>>(d_c, d_a, scalar, N);
    cudaThreadSynchronize();
    scalar = 3.0f;
    STREAM_Triad<cudaGrid, dimBlock>>(d_b, d_a, scalar, N);
    cudaFree(d_a);
    cudaFree(d_b);
    cudaFree(d_c);
}
    
```

- Chapel's support for parallelism and locality supports...
 - ...concise, clear, yet portable benchmarks
 - ...user applications that are productive and scalable

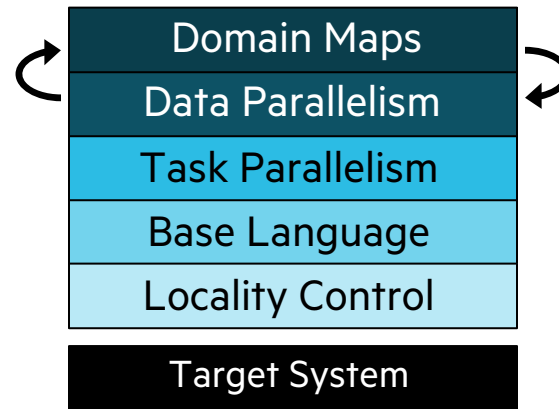


INTRODUCTION TO CHAPEL

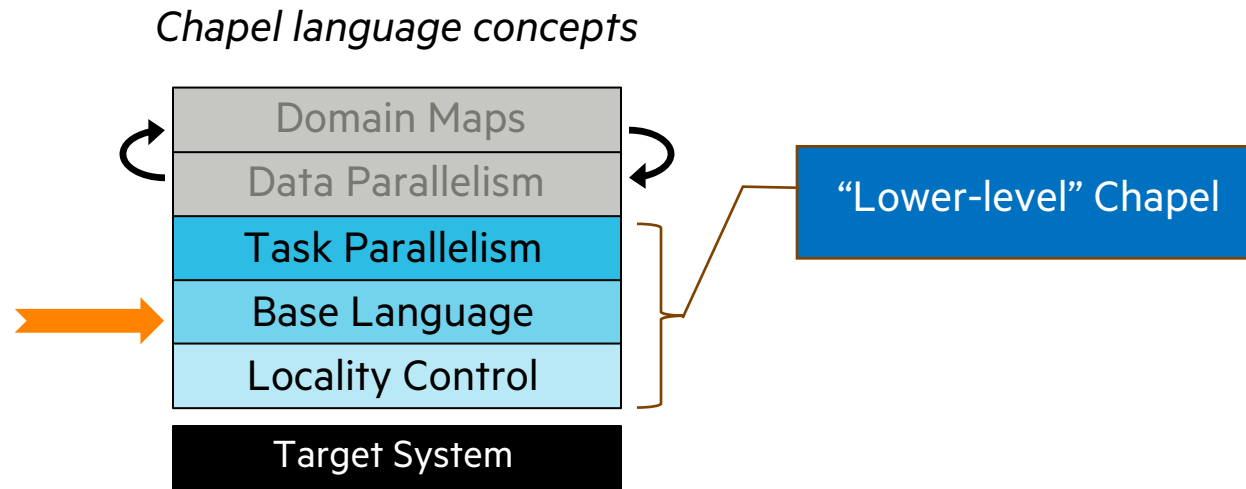


CHAPEL FEATURE AREAS

Chapel language concepts



BASE LANGUAGE



FIBONACCI ITERATION

fib.chpl

```
config const n = 10;

for f in fib(n) do
  writeln(f);

iter fib(x) {
  var current = 0,
      next = 1;

  for i in 1..x {
    yield current;
    current += next;
    current <=> next;
  }
}
```

```
prompt> chpl fib.chpl
prompt>
```

FIBONACCI ITERATION

fib.chpl

```
config const n = 10;

for f in fib(n) do
  writeln(f);

iter fib(x) {
  var current = 0,
      next = 1;

  for i in 1..x {
    yield current;
    current += next;
    current <=> next;
  }
}
```

Drive this loop
by invoking fib(n)

```
prompt> chpl fib.chpl
prompt> ./fib
```

FIBONACCI ITERATION

fib.chpl

```
config const n = 10;  
  
for f in fib(n) do  
  writeln(f);  
  
iter fib(x) {  
  var current = 0,  
      next = 1;  
  
  for i in 1..x {  
    yield current;  
    current += next;  
    current <=> next;  
  }  
}
```

Execute the loop's body
for that value

'yield' this expression back
to the loop's index variable

```
prompt> chpl fib.chpl  
prompt> ./fib  
0
```

FIBONACCI ITERATION

fib.chpl

```
config const n = 10;  
  
for f in fib(n) do  
  writeln(f);  
  
iter fib(x) {  
  var current = 0,  
      next = 1;  
  
  for i in 1..x {  
    yield current;  
    current += next;  
    current <=> next;  
  }  
}
```

Execute the loop's body
for that value

Then continue the iterator
from where it left off

Repeating until we fall
out of it (or return)

```
prompt> chpl fib.chpl  
prompt> ./fib  
0  
1  
1  
2  
3  
5  
8  
13  
21  
34
```


FIBONACCI ITERATION

fib.chpl

```
config const n = 10;

for f in fib(n) do
  writeln(f);

iter fib(x) {
  var current = 0,
      next = 1;

  for i in 1..x {
    yield current;
    current += next;
    current <=> next;
  }
}
```

Config[urable] declarations
support command-line overrides

```
prompt> chpl fib.chpl
prompt> ./fib --n=1000
0
1
1
2
3
5
8
13
21
34
55
89
144
233
377
...
```

FIBONACCI ITERATION

fib.chpl

```
config const n = 10;

for f in fib(n) do
  writeln(f);

iter fib(x) {
  var current = 0,
      next = 1;

  for i in 1..x {
    yield current;
    current += next;
    current <=> next;
  }
}
```

Static type inference for:

- constants / variables
- arguments
- return types

Explicit typing also supported

```
prompt> chpl fib.chpl
prompt> ./fib --n=1000
0
1
1
2
3
5
8
13
21
34
55
89
144
233
377
...
```

FIBONACCI ITERATION

fib.chpl

```
config const n: int = 10;

for f in fib(n) do
  writeln(f);

iter fib(x: int): int {
  var current: int = 0,
      next: int = 1;

  for i in 1..x {
    yield current;
    current += next;
    current <=> next;
  }
}
```

Explicit typing also supported

```
prompt> chpl fib.chpl
prompt> ./fib --n=1000
0
1
1
2
3
5
8
13
21
34
55
89
144
233
377
...
```

FIBONACCI ITERATION

fib.chpl

```
config const n = 10;

for (i,f) in zip(0..<n, fib(n)) do
  writeln("fib #", i, " is ", f);

iter fib(x) {
  var current = 0,
      next = 1;

  for i in 1..x {
    yield current;
    current += next;
    current <=> next;
  }
}
```

Zippered
iteration

```
prompt> chpl fib.chpl
prompt> ./fib --n=1000
fib #0 is 0
fib #1 is 1
fib #2 is 1
fib #3 is 2
fib #4 is 3
fib #5 is 5
fib #6 is 8
fib #7 is 13
fib #8 is 21
fib #9 is 34
fib #10 is 55
fib #11 is 89
fib #12 is 144
fib #13 is 233
fib #14 is 377
...
```

FIBONACCI ITERATION

fib.chpl

```
config const n = 10;

for (i,f) in zip(0..
```

Range types
and operators

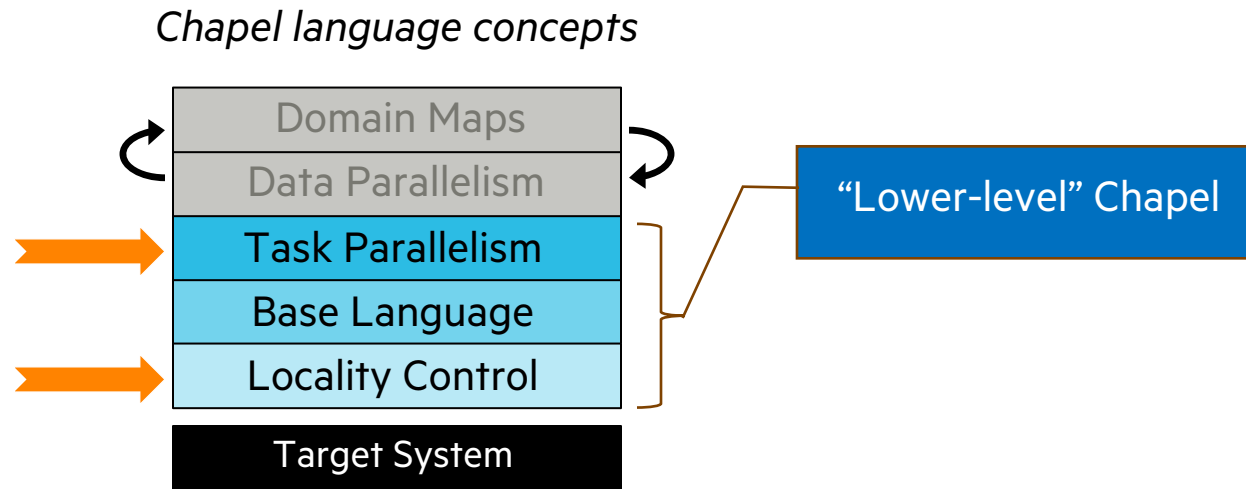
```
prompt> chpl fib.chpl
prompt> ./fib --n=1000
fib #0 is 0
fib #1 is 1
fib #2 is 1
fib #3 is 2
fib #4 is 3
fib #5 is 5
fib #6 is 8
fib #7 is 13
fib #8 is 21
fib #9 is 34
fib #10 is 55
fib #11 is 89
fib #12 is 144
fib #13 is 233
fib #14 is 377
...
```

OTHER BASE LANGUAGE FEATURES

- **Various basic types:** bool(w), int(w), uint(w), real(w), imag(w), complex(w), enums, tuples
- **Object-oriented programming**
 - Value-based records (like C structs supporting methods, generic fields, etc.)
 - Reference-based classes (somewhat like Java classes or C++ pointers-to-classes)
 - Nilable vs. non-nilable variants
 - Memory-management strategies (shared, owned, borrowed, unmanaged)
 - Lifetime checking
- **Error-handling**
- **Generic programming / polymorphism**
- **Compile-time meta-programming**
- **Modules** (supporting namespaces)
- **Procedure overloading / filtering**
- **Arguments:** default values, intents, name-based matching, type queries
- and more...



TASK PARALLELISM AND LOCALITY CONTROL

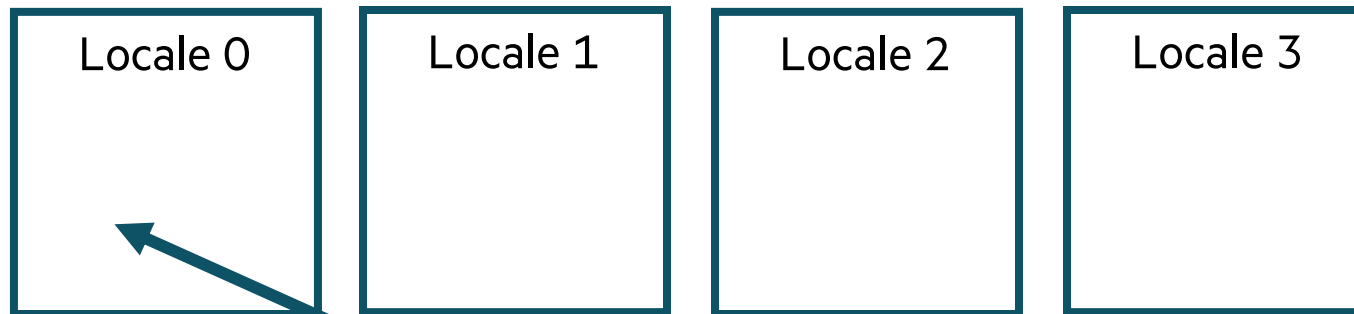


THE LOCALE: CHAPEL'S KEY FEATURE FOR LOCALITY

- *locale*: a unit of the target architecture that can run tasks and store variables
 - Think “compute node” on a typical HPC system

```
prompt> ./myChapelProgram --numLocales=4 # or '-nl 4'
```

Locales array :



User's program starts running as a single task on locale 0



TASK-PARALLEL “HELLO WORLD”

helloTaskPar.chpl

```
const numTasks = here.maxTaskPar;  
coforall tid in 1..numTasks do  
    writef("Hello from task %n of %n on %s\n",  
          tid, numTasks, here.name);
```

TASK-PARALLEL “HELLO WORLD”

helloTaskPar.chpl

```
const numTasks = here.maxTaskPar;  
coforall tid in 1..numTasks do  
  writef("Hello from task %n of %n on %s\n",  
        tid, numTasks, here.name);
```

‘here’ refers to the locale on which this code is currently running

how many parallel tasks can my locale run at once?

what’s my locale’s name?



TASK-PARALLEL “HELLO WORLD”

helloTaskPar.chpl

```
const numTasks = here.maxTaskPar;  
coforall tid in 1..numTasks do  
    writef("Hello from task %n of %n on %s\n",  
          tid, numTasks, here.name);
```

a 'coforall' loop executes each iteration as an independent task

```
prompt> chpl helloTaskPar.chpl  
prompt> ./helloTaskPar  
Hello from task 1 of 4 on n1032  
Hello from task 4 of 4 on n1032  
Hello from task 3 of 4 on n1032  
Hello from task 2 of 4 on n1032
```



TASK-PARALLEL “HELLO WORLD”

helloTaskPar.chpl

```
const numTasks = here.maxTaskPar;  
coforall tid in 1..numTasks do  
    writef("Hello from task %n of %n on %s\n",  
          tid, numTasks, here.name);
```

```
prompt> chpl helloTaskPar.chpl  
prompt> ./helloTaskPar  
Hello from task 1 of 4 on n1032  
Hello from task 4 of 4 on n1032  
Hello from task 3 of 4 on n1032  
Hello from task 2 of 4 on n1032
```

So far, this is a shared-memory program

Nothing refers to remote locales,
explicitly or implicitly

TASK-PARALLEL “HELLO WORLD”

helloTaskPar.chpl

```
const numTasks = here.maxTaskPar;  
coforall tid in 1..numTasks do  
    writef("Hello from task %n of %n on %s\n",  
          tid, numTasks, here.name);
```

TASK-PARALLEL “HELLO WORLD” (DISTRIBUTED VERSION)

helloTaskPar.chpl

```
coforall loc in Locales {  
  on loc {  
    const numTasks = here.maxTaskPar;  
    coforall tid in 1..numTasks do  
      writef("Hello from task %n of %n on %s\n",  
            tid, numTasks, here.name);  
  }  
}
```

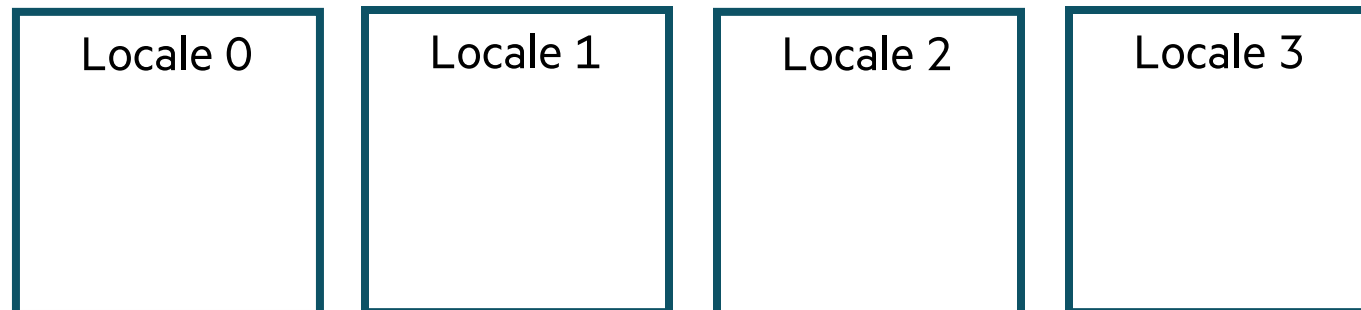
TASK-PARALLEL “HELLO WORLD” (DISTRIBUTED VERSION)

helloTaskPar.chpl

```
coforall loc in Locales {  
  on loc {  
    const numTasks = here.maxTaskPar;  
    coforall tid in 1..numTasks do  
      writef("Hello from task %n of %n on %s\n",  
            tid, numTasks, here.name);  
  }  
}
```

the array of locales we're running on
(introduced a few slides back)

Locales array:



TASK-PARALLEL “HELLO WORLD” (DISTRIBUTED VERSION)

helloTaskPar.chpl

```
coforall loc in Llocales {  
  on loc {  
    const numTasks = here.maxTaskPar;  
    coforall tid in 1..numTasks do  
      writef("Hello from task %n of %n on %s\n",  
            tid, numTasks, here.name);  
  }  
}
```

create a task per locale
on which the program is running

have each task run 'on' its locale

then print a message per core,
as before

```
prompt> chpl helloTaskPar.chpl  
prompt> ./helloTaskPar -numLocales=4  
Hello from task 1 of 4 on n1032  
Hello from task 4 of 4 on n1032  
Hello from task 1 of 4 on n1034  
Hello from task 2 of 4 on n1032  
Hello from task 1 of 4 on n1033  
Hello from task 3 of 4 on n1034  
Hello from task 1 of 4 on n1035  
...
```


TASK-PARALLEL “HELLO WORLD” (DISTRIBUTED VERSION)

helloTaskPar.chpl

```
coforall loc in Locales {  
  on loc {  
    const numTasks = here.maxTaskPar;  
    coforall tid in 1..numTasks do  
      writef("Hello from task %n of %n on %s\n",  
            tid, numTasks, here.name);  
  }  
}
```

PARALLELISM AND LOCALITY ARE ORTHOGONAL IN CHAPEL

- This is a parallel, but local program:

```
coforall i in 1..msgs do
  writeln("Hello from task ", i);
```

- This is a distributed, but serial program:

```
writeln("Hello from locale 0!");
on Locales[1] do writeln("Hello from locale 1!");
on Locales[2] {
  writeln("Hello from locale 2!");
  on Locales[0] do writeln("Hello from locale 0!");
}
writeln("Back on locale 0");
```

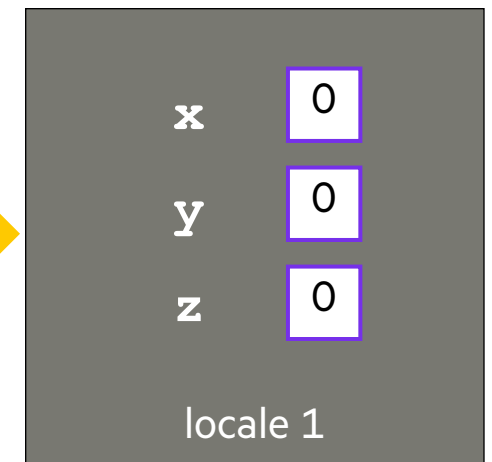
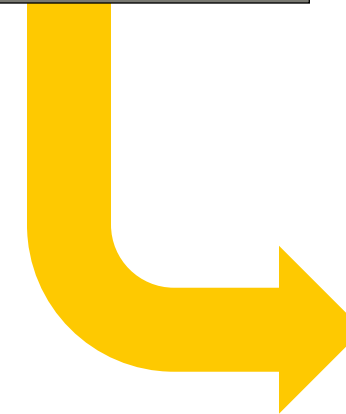
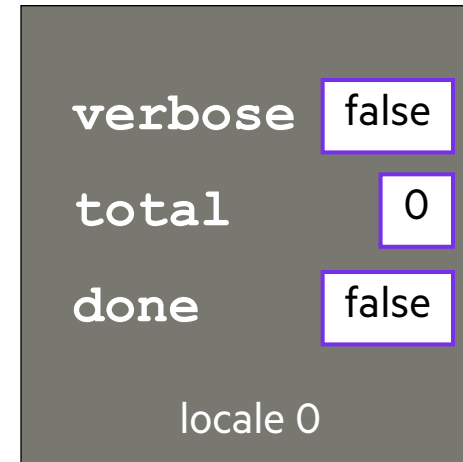
- This is a distributed parallel program:

```
coforall i in 1..msgs do
  on Locales[i%numLocales] do
    writeln("Hello from task ", i, " running on locale ", here.id);
```

VARIABLES ARE ALLOCATED LOCALLY TO WHERE THE TASK IS RUNNING

onClause.chpl

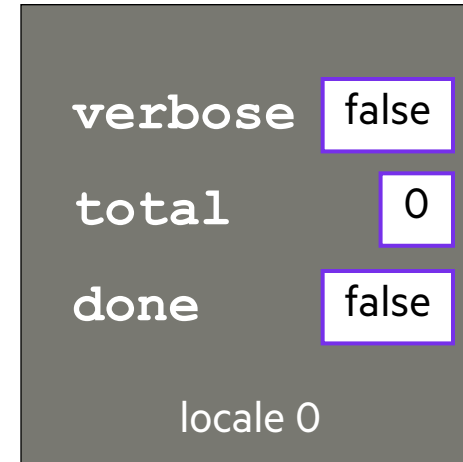
```
config const verbose = false;  
var total = 0,  
    done = false;  
  
...  
  
on Locales [1] {  
    var x, y, z: int;  
    ...  
}
```



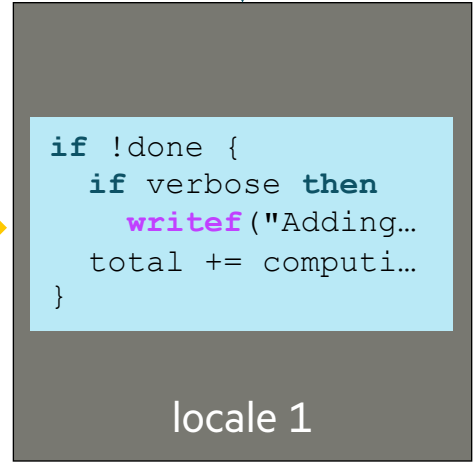
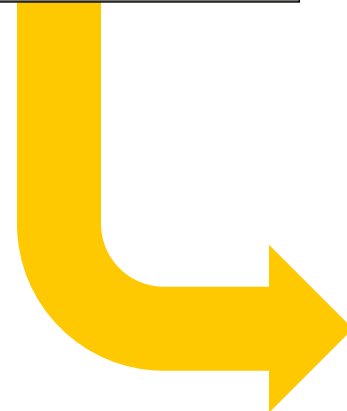
CODE CAN REFER TO VISIBLE VARIABLES, EVEN WHEN THEY'RE REMOTE

onClause.chpl

```
config const verbose = false;
var total = 0,
    done = false;
...
on Locales [1] {
  if !done {
    if verbose then
      writef("Adding locale 1's contribution");
    total += computeMyContribution();
  }
}
```



code runs on locale 1,
but refers to values
stored on locale 0



SIDEBAR: CHAPEL'S DECEPTIVE SIMPLICITY

Chapel resembles traditional programming enough that it's easy to forget how roundabout SPMD can be:

- Need to do something on just one node?
 - **Chapel:** OK, just do it
 - **SPMD:** Make sure you're only doing it in one process

toy.chpl

```
proc main() {  
  var x = stdin.read(int);  
  writeln("Hello!");  
  
  coforall loc in Locales do  
    on loc do  
      writeln(loc.id * x);  
  
  writeln("Bye!");  
}
```

toy-SPMD.chpl

```
proc main() {  
  var x: int;  
  if myProc() == 0 {  
    x = stdin.read(int);  
    writeln("Hello!");  
  }  
  
  broadcastAll(x, fromLocale=0);  
  
  writeln(myProc() * x);  
  
  barrierAll();  
  
  if myProc() == 0 then  
    writeln("Bye!");  
}
```

SIDEBAR: CHAPEL'S DECEPTIVE SIMPLICITY

Chapel resembles traditional programming enough that it's easy to forget how roundabout SPMD can be:

- Want to ensure one thing finishes before the next?
 - **Chapel:** Typically happens through sequential ordering
 - **SPMD:** Defensively ensure nobody gets too far ahead

toy.chpl

```
proc main() {  
  var x = stdin.read(int);  
  writeln("Hello!");  
  
  coforall loc in Locales do  
    on loc do  
      writeln(loc.id * x);  
  
  writeln("Bye!");  
}
```

toy-SPMD.chpl

```
proc main() {  
  var x: int;  
  if myProc() == 0 {  
    x = stdin.read(int);  
    writeln("Hello!");  
  }  
  
  broadcastAll(x, fromLocale=0);  
  
  writeln(myProc() * x);  
  
  barrierAll();  
  
  if myProc() == 0 then  
    writeln("Bye!");  
}
```

SIDEBAR: CHAPEL'S DECEPTIVE SIMPLICITY

Chapel resembles traditional programming enough that it's easy to forget how roundabout SPMD can be:

- Want to refer to a remote variable?
 - **Chapel:** Is it in your lexical scope? Just name it!
 - **SPMD:** Insert communication, potentially in both the source and destination processes

toy.chpl

```
proc main() {
  var x = stdin.read(int);
  writeln("Hello!");

  coforall loc in Locales do
    on loc do
      writeln(loc.id * x);

  writeln("Bye!");
}
```

toy-SPMD.chpl

```
proc main() {
  var x: int;
  if myProc() == 0 {
    x = stdin.read(int);
    writeln("Hello!");
  }

  broadcastAll(x, fromLocale=0);

  writeln(myProc() * x);

  barrierAll();

  if myProc() == 0 then
    writeln("Bye!");
}
```

SIDEBAR: CHAPEL'S DECEPTIVE SIMPLICITY

Chapel resembles traditional programming enough that it's easy to forget how roundabout SPMD can be:

- Need some additional parallelism?
 - **Chapel:** we have features for that, like `coforall`, logically independent of hardware resources
 - **SPMD:** Umm... Well, I suppose you could mix in OpenMP, Pthreads, or CUDA...

toy.chpl

```
proc main() {  
  var x = stdin.read(int);  
  writeln("Hello!");  
  
  forall loc in Locales do  
    on loc do  
      writeln(loc.id * x);  
  
  writeln("Bye!");  
}
```

toy-SPMD.chpl

```
proc main() {  
  var x: int;  
  if myProc() == 0 {  
    x = stdin.read(int);  
    writeln("Hello!");  
  }  
  
  broadcastAll(x, fromLocale=0);  
  
  writeln(myProc() * x);  
  
  barrierAll();  
  
  if myProc() == 0 then  
    writeln("Bye!");  
}
```


SIDEBAR: CHAPEL'S DECEPTIVE SIMPLICITY

Chapel resembles traditional programming enough that it's easy to forget how roundabout SPMD can be:

- And of course, if what you really want is SPMD, Chapel can do that as well...

toy.chpl

```
proc main() {
  var x = stdin.read(int);
  writeln("Hello!");

  coforall loc in Locales do
    on loc do
      writeln(loc.id * x);

  writeln("Bye!");
}
```

toy-SPMD.chpl

```
proc main() {
  var x: int;
  if myProc() == 0 {
    x = stdin.read(int);
    writeln("Hello!");
  }

  broadcastAll(x, fromLocale=0);

  writeln(myProc() * x);

  barrierAll();

  if myProc() == 0 then
    writeln("Bye!");
}
```

OTHER TASK PARALLEL FEATURES

- **begin / cobegin statements:** the two other ways of creating tasks

```
begin stmt; // fire off an asynchronous task to run 'stmt'
```

```
cobegin { // fire off a task for each of 'stmt1', 'stmt2', ...  
  stmt1;  
  stmt2;  
  stmt3;  
  ...  
} // wait here for these tasks to complete before proceeding
```

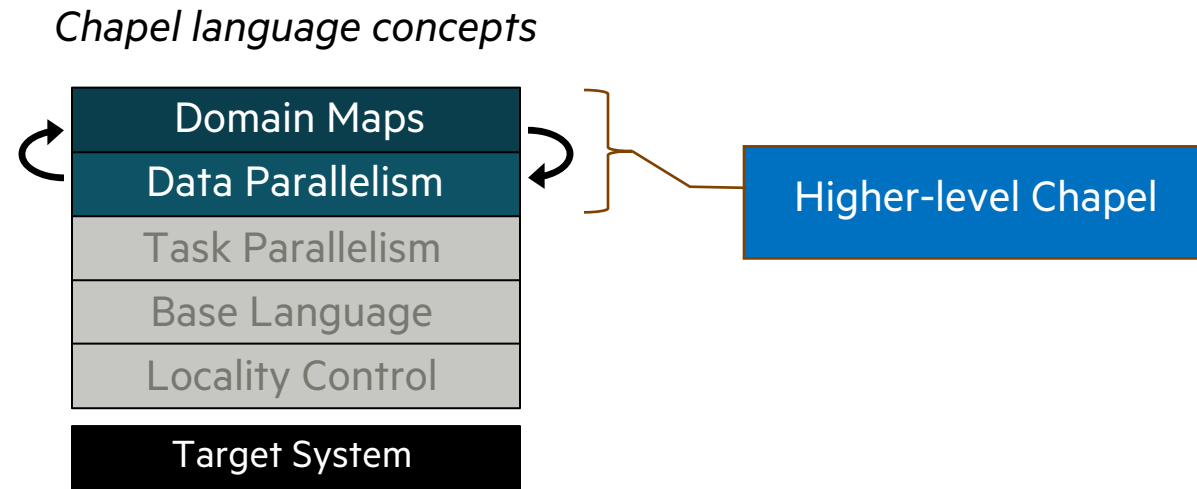
- **atomic / synchronized variables:** types for safe data sharing & coordination between tasks

```
var sum: atomic int; // supports various atomic methods like .add(), .compareExchange(), ...  
var cursor: sync int; // stores a full/empty bit governing reads/writes, supporting .readEFO(), .writeEFO()
```

- **task intents / task-private variables:** control how variables and tasks relate

```
coforall i in 1..nitems with (ref x, + reduce y, var z: int) { ... }
```

DATA PARALLELISM AND DOMAIN MAPS



DATA-PARALLEL ARRAY FILL

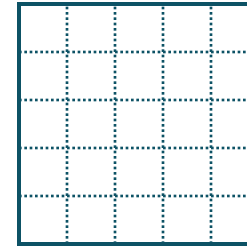
fillArray.chpl

```
config const n = 1000;  
  
const D = {1..n, 1..n};  
  
var A: [D] real;  
  
forall (i,j) in D do  
    A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```

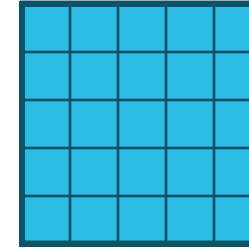
DATA-PARALLEL ARRAY FILL

fillArray.chpl

```
config const n = 1000;  
  
const D = {1..n, 1..n};  
  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



D



A

declare a domain, a first-class index set

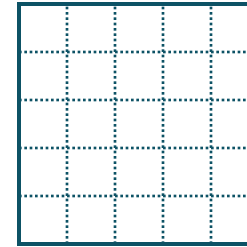
declare an array over that domain



DATA-PARALLEL ARRAY FILL

fillArray.chpl

```
config const n = 1000;  
  
const D = {1..n, 1..n};  
  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



D

1.1	1.3	1.5	1.5	1.9
2.1	2.3	2.5	2.7	2.9
3.1	3.3	3.5	3.7	3.9
4.1	4.3	4.5	4.7	4.9
5.1	5.3	5.5	5.7	5.9

A

declare a domain, a first-class index set

declare an array over that domain

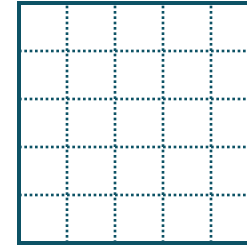
iterate over the domain's indices in parallel,
assigning to the corresponding array elements



DATA-PARALLEL ARRAY FILL

fillArray.chpl

```
config const n = 1000;  
  
const D = {1..n, 1..n};  
  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



A 5x5 grid representing the domain D, with dashed lines forming the grid structure.

D

1.1	1.3	1.5	1.7	1.9
2.1	2.3	2.5	2.7	2.9
3.1	3.3	3.5	3.7	3.9
4.1	4.3	4.5	4.7	4.9
5.1	5.3	5.5	5.7	5.9

A

```
prompt> chpl dataParallel.chpl  
prompt> ./dataParallel --n=5  
1.1 1.3 1.5 1.7 1.9  
2.1 2.3 2.5 2.7 2.9  
3.1 3.3 3.5 3.7 3.9  
4.1 4.3 4.5 4.7 4.9  
5.1 5.3 5.5 5.7 5.9
```

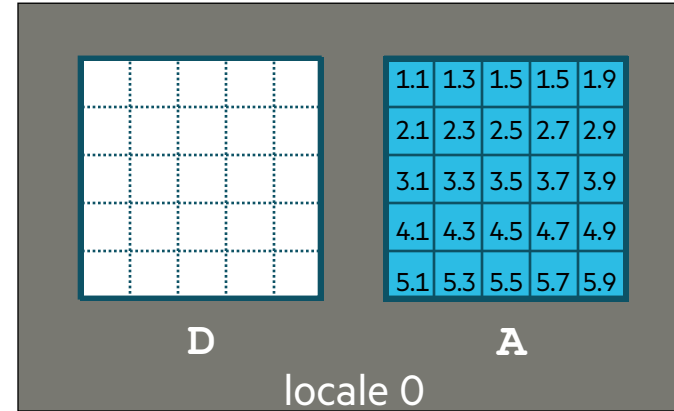
So far, this is a shared-memory program

Nothing refers to remote locales,
explicitly or implicitly

DATA-PARALLEL ARRAY FILL

fillArray.chpl

```
config const n = 1000;  
  
const D = {1..n, 1..n};  
  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



```
prompt> chpl dataParallel.chpl  
prompt> ./dataParallel --n=5  
1.1 1.3 1.5 1.7 1.9  
2.1 2.3 2.5 2.7 2.9  
3.1 3.3 3.5 3.7 3.9  
4.1 4.3 4.5 4.7 4.9  
5.1 5.3 5.5 5.7 5.9
```

So far, this is a shared-memory program

Nothing refers to remote locales,
explicitly or implicitly

DATA-PARALLEL ARRAY FILL

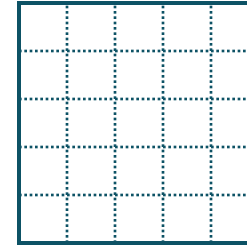
fillArray.chpl

```
config const n = 1000;  
  
const D = {1..n, 1..n};  
  
var A: [D] real;  
  
forall (i,j) in D do  
    A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```

DATA-PARALLEL ARRAY FILL (DISTRIBUTED VERSION)

fillArray.chpl

```
use CyclicDist;  
  
config const n = 1000;  
  
const D = {1..n, 1..n}  
         dmapped Cyclic(startIdx = (1,1));  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



D

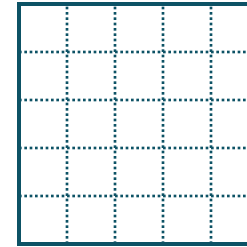
1.1	1.3	1.5	1.5	1.9
2.1	2.3	2.5	2.7	2.9
3.1	3.3	3.5	3.7	3.9
4.1	4.3	4.5	4.7	4.9
5.1	5.3	5.5	5.7	5.9

A

DATA-PARALLEL ARRAY FILL (DISTRIBUTED VERSION)

fillArray.chpl

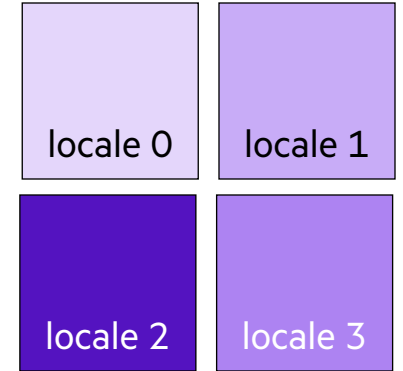
```
use CyclicDist;  
  
config const n = 1000;  
  
const D = {1..n, 1..n}  
         dmapped Cyclic(startIdx = (1,1));  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



D

1.1	1.3	1.5	1.5	1.9
2.1	2.3	2.5	2.7	2.9
3.1	3.3	3.5	3.7	3.9
4.1	4.3	4.5	4.7	4.9
5.1	5.3	5.5	5.7	5.9

A



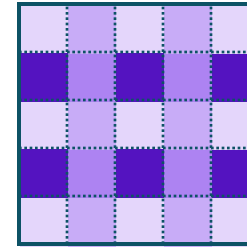
apply a domain map, specifying how to implement..
...the domain's indices,
...the array's elements,
...the loop's iterations,
...on the program's locales



DATA-PARALLEL ARRAY FILL (DISTRIBUTED VERSION)

fillArray.chpl

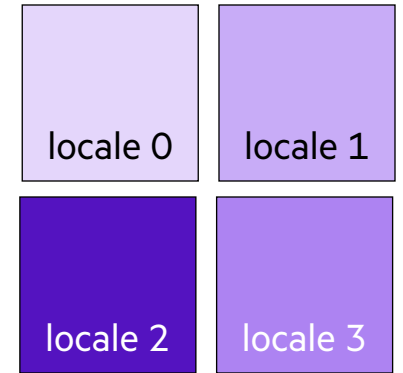
```
use CyclicDist;  
  
config const n = 1000;  
  
const D = {1..n, 1..n}  
         dmapped Cyclic(startIdx = (1,1));  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



D

1.1	1.3	1.5	1.5	1.9
2.1	2.3	2.5	2.7	2.9
3.1	3.3	3.5	3.7	3.9
4.1	4.3	4.5	4.7	4.9
5.1	5.3	5.5	5.7	5.9

A

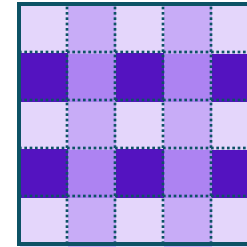


apply a domain map, specifying how to implement..
...the domain's indices,
...the array's elements,
...the loop's iterations,
...on the program's locales

DATA-PARALLEL ARRAY FILL (DISTRIBUTED VERSION)

fillArray.chpl

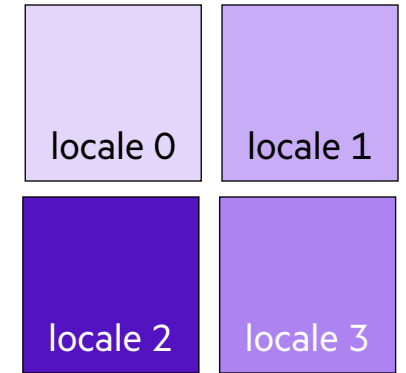
```
use CyclicDist;  
  
config const n = 1000;  
  
const D = {1..n, 1..n}  
         dmapped Cyclic(startIdx = (1,1));  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



D

1.1	1.3	1.5	1.5	1.9
2.1	2.3	2.5	2.7	2.9
3.1	3.3	3.5	3.7	3.9
4.1	4.3	4.5	4.7	4.9
5.1	5.3	5.5	5.7	5.9

A

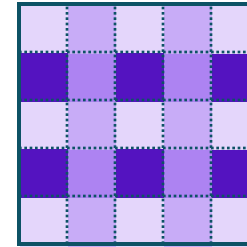


```
prompt> chpl dataParallel.chpl  
prompt> ./dataParallel --n=5  
1.1 1.3 1.5 1.7 1.9  
2.1 2.3 2.5 2.7 2.9  
3.1 3.3 3.5 3.7 3.9  
4.1 4.3 4.5 4.7 4.9  
5.1 5.3 5.5 5.7 5.9
```

DATA-PARALLEL ARRAY FILL (DISTRIBUTED VERSION)

fillArray.chpl

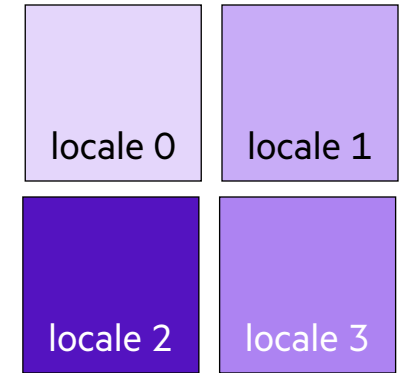
```
use CyclicDist;  
  
config const n = 1000;  
  
const D = {1..n, 1..n}  
         dmapped Cyclic(startIdx = (1,1));  
var A: [D] real;  
  
forall (i,j) in D do  
  A[i,j] = i + (j - 0.5)/n;  
  
writeln(A);
```



D

1.1	1.3	1.5	1.7	1.9
2.1	2.3	2.5	2.7	2.9
3.1	3.3	3.5	3.7	3.9
4.1	4.3	4.5	4.7	4.9
5.1	5.3	5.5	5.7	5.9

A



```
prompt> chpl dataParallel.chpl  
prompt> ./dataParallel --n=5 --numLocales=4  
1.1 1.3 1.5 1.7 1.9  
2.1 2.3 2.5 2.7 2.9  
3.1 3.3 3.5 3.7 3.9  
4.1 4.3 4.5 4.7 4.9  
5.1 5.3 5.5 5.7 5.9
```

DATA-PARALLEL ARRAY FILL (DISTRIBUTED VERSION)

fillArray.chpl

```
use CyclicDist;

config const n = 1000;

const D = {1..n, 1..n}
         dmapped Cyclic(startIdx = (1,1));
var A: [D] real;

forall (i,j) in D do
  A[i,j] = i + (j - 0.5)/n;

writeln(A);
```

SPECTRUM OF CHAPEL FOR-LOOP STYLES

for loop: each iteration is executed serially by the current task

- predictable execution order, similar to conventional languages

foreach loop: all iterations executed by the current task, but in no specific order

- a candidate for vectorization, SIMD execution on GPUs

forall loop: all iterations are executed by one or more tasks in no specific order

- implemented using one or more tasks, locally or distributed, as determined by the iterand expression

```
forall i in 1..n do ... // forall loops over ranges use local tasks only
forall (i,j) in {1..n, 1..n} do ... // ditto for local domains...
forall elem in myLocArr do ... // ...and local arrays
forall elem in myDistArr do ... // distributed arrays use tasks on each locale owning part of the array
forall i in myParIter(...) do ... // you can also write your own iterators that use the policy you want
```

coforall loop: each iteration is executed concurrently by a distinct task

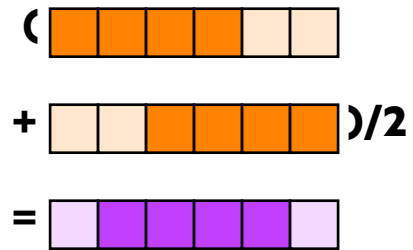
- explicit parallelism; supports synchronization between iterations (tasks)



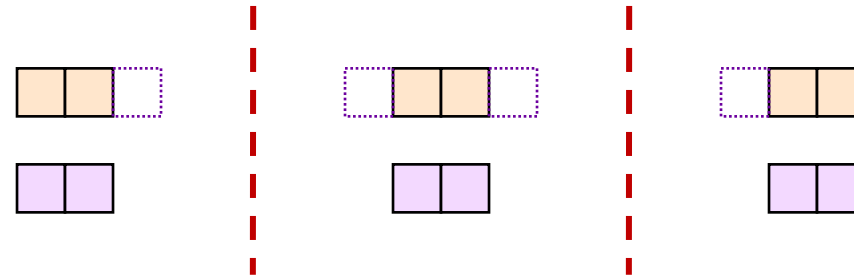
CHAPEL'S GLOBAL-VIEW OF DATA-PARALLELISM VS. SPMD

- “Apply a 3-point stencil to a vector”

Global-View



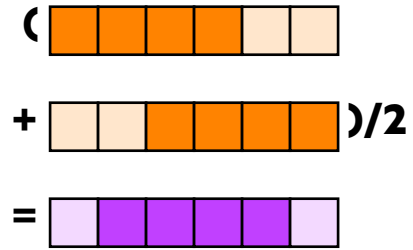
SPMD



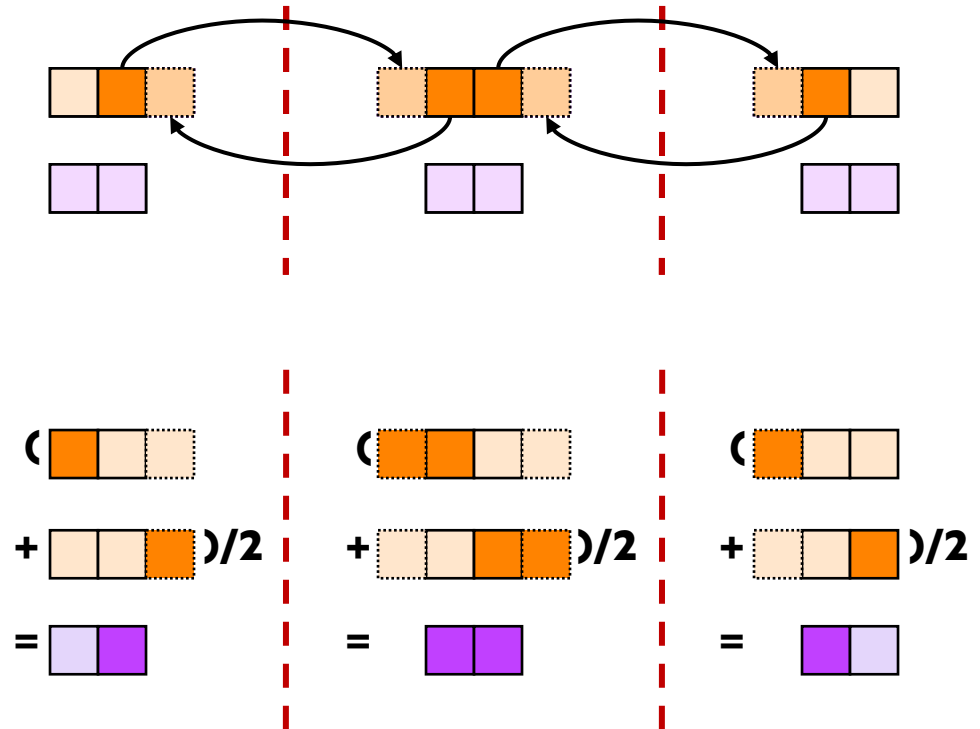
CHAPEL'S GLOBAL-VIEW OF DATA-PARALLELISM VS. SPMD

- “Apply a 3-point stencil to a vector”

Global-View




SPMD



CHAPEL'S GLOBAL-VIEW VS. SPMD

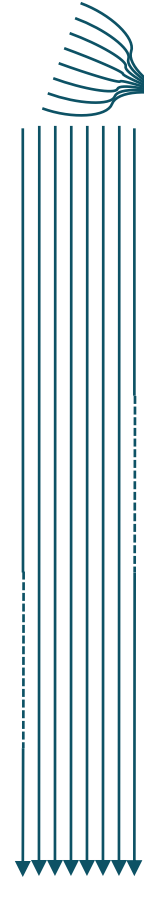
- “Apply a 3-point stencil to a vector”

Global-View



```
proc main() {  
  var n = 1000;  
  var A, B: [1..n] real;  
  
  forall i in 2..n-1 do  
    B[i] = (A[i-1] + A[i+1])/2;  
  }  
}
```

SPMD (MPI-style)



```
proc main() {  
  var n = 1000;  
  var p = numProcs(),  
      me = myProc(),  
      myN = n/p,  
  var A, B: [0..myN+1] real;  
  
  if (me < p-1) {  
    send(me+1, A[myN]);  
    recv(me+1, A[myN+1]);  
  }  
  if (me > 0) {  
    send(me-1, A[1]);  
    recv(me-1, A[0]);  
  }  
  forall i in 1..myN do  
    B[i] = (A[i-1] + A[i+1])/2;  
  }  
}
```

Bug: Refers to uninitialized values at ends of A




CHAPEL'S GLOBAL-VIEW VS. SPMD

- “Apply a 3-point stencil to a vector”

Global-View


```
proc main() {  
  var n = 1000;  
  var A, B: [1..n] real;  
  
  forall i in 2..n-1 do  
    B[i] = (A[i-1] + A[i+1])/2;  
  }  
}
```



Communication becomes geometrically more complex for higher-dimensional arrays

SPMD (MPI-style)

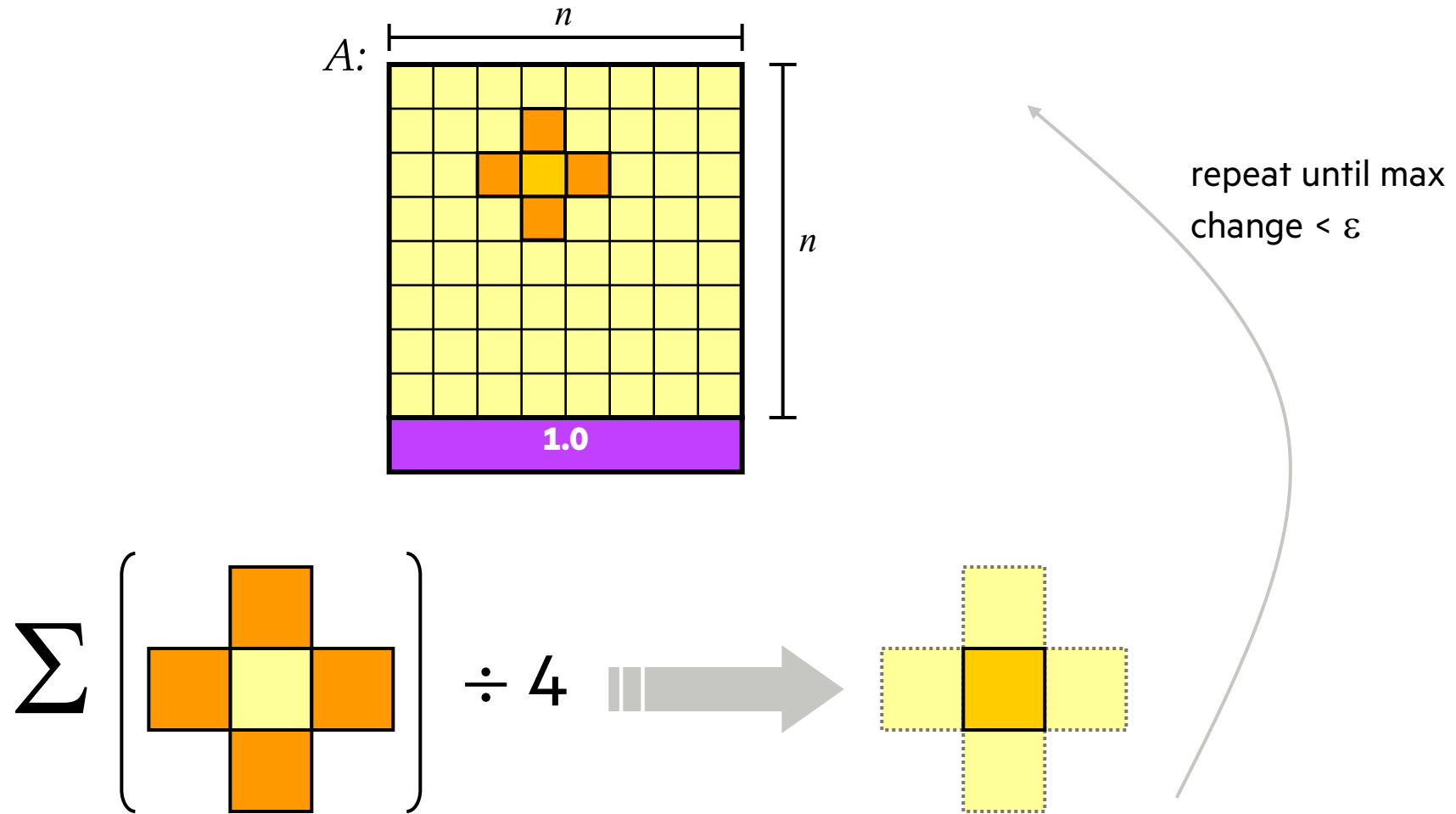
```
proc main() {  
  var n = 1000;  
  var p = numProcs(),  
      me = myProc(),  
      myN = n/p,  
      myLo = 1,  
      myHi = myN;  
  var A, B: [0..myN+1] real;  
  
  if (me < p-1) {  
    send(me+1, A[myN]);  
    recv(me+1, A[myN+1]);  
  } else  
    myHi = myN-1;  
  if (me > 0) {  
    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;  
  }  
}
```



Assumes p evenly divides n



JACOBI ITERATION IN PICTURES



JACOBI ITERATION IN CHAPEL

```
config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1},
       D = AllInds[1..n, 1..n],
       LastRow = AllInds[n+1..n+1, ..];

var A, Temp: [AllInds] real;

A[LastRow] = 1.0;

do {
  forall (i,j) in D do
    Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;

  const delta = max reduce abs(A[D] - Temp[D]);
  A[D] = Temp[D];
} while delta > epsilon;

writeln(A);
```

JACOBI ITERATION IN CHAPEL

```
config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1},
      D = AllInds[1..n, 1..n],
      LastRow = AllInds[n+1..n+1, ..];

var A, Temp: [AllInds] real;

A[LastRow] = 1.0;

do {
  forall (i,j) in D do
    Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;

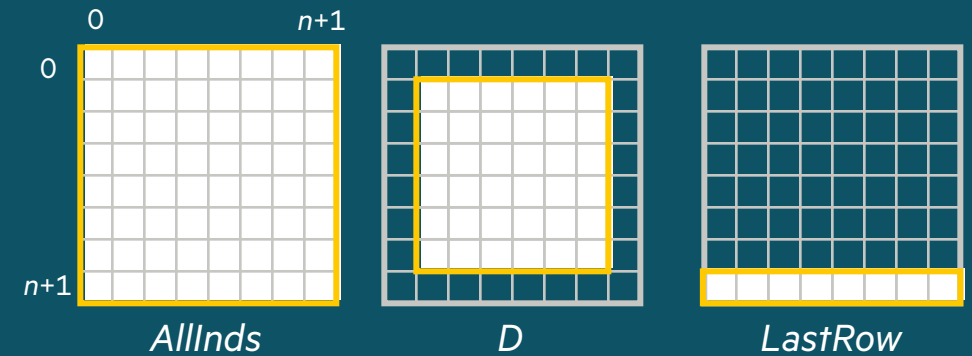
  const delta = max reduce abs (A[D] - Temp[D]);
  A[D] = Temp[D];
} while delta > epsilon;

writeln(A);
```

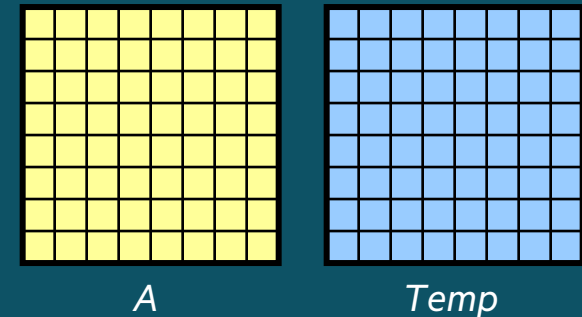
Declare configurable problem size and termination condition

Declare domains

- Slicing one domain with another computes the intersection



Declare arrays



JACOBI ITERATION IN CHAPEL

```
config const n = 6, epsilon = 0.01;
```

```
const AllInds = {0..n+1, 0..n+1},  
      D = AllInds[1..n, 1..n],  
      LastRow = AllInds[n+1..n+1, ..];
```

```
var A, Temp: [AllInds] real;
```

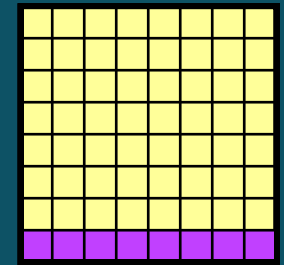
```
A[LastRow] = 1.0;
```

```
do {  
  forall (i,j) in D do  
    Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;  
  
  const delta = max reduce abs (A[D] - Temp[D]);  
  A[D] = Temp[D];  
} while delta > epsilon;
```

```
writeln(A);
```

Set Explicit Boundary Condition

- indexing by a domain refers to the subarray in question
- scalar values are “promoted” when assigned to arrays
- “whole-array” operations like this are implicitly parallel



A

JACOBI ITERATION IN CHAPEL

```
config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1},
      D = AllInds[1..n, 1..n],
      LastRow = AllInds[n+1..n+1, ..];

var A, Temp: [AllInds] real;

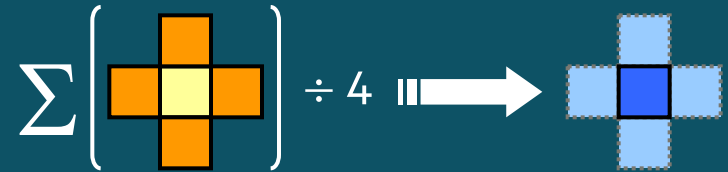
A[LastRow] = 1.0;

do {
  forall (i,j) in D do
    Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;

  const delta = max reduce abs (A[D] - Temp[D]);
  A[D] = Temp[D];
} while delta > epsilon;

writeln(A);
```

Compute 5-point stencil



JACOBI ITERATION IN CHAPEL

```
config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1},
       D = AllInds[1..n, 1..n],
       LastRow = AllInds[n+1..n+1, ..];
```

```
var A, Temp: [AllInds] real;
```

```
A[LastRow] = 1.0;
```

```
do {
  forall (i,j) in D do
    Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;

  const delta = max reduce abs(A[D] - Temp[D]);
  A[D] = Temp[D];
} while delta > epsilon;
```

```
writeln(A);
```

Compute maximum change

- *op* **reduce** *expr* \Rightarrow collapse aggregate expression '*expr*' to a scalar using '*op*'
- `abs()` and '-' are scalar operations; calling them with array arguments results in parallel evaluation
- no temporary arrays are created when evaluating this statement



SIDEBAR: PROMOTION OF SCALAR SUBROUTINES

- Any function or operator that takes scalar arguments can be called with array expressions instead

```
proc foo(x: real, y: real, z: real) {  
  return x**y + 10*c;  
}
```

- Interpretation is similar to that of a zippered forall loop, thus:

```
C = foo(A, 2, B);
```

is equivalent to:

```
forall (c, a, b) in zip(C, A, B) do  
  c = foo(a, 2, b);
```

as is:

```
C = A**2 + 10*c;
```

- So, in the Jacobi computation,

```
abs(A[D] - Temp[D]); == forall (a,t) in zip(A[D], Temp[D]) do abs(a - t);
```

JACOBI ITERATION IN CHAPEL

```
config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1},
      D = AllInds[1..n, 1..n],
      LastRow = AllInds[n+1..n+1, ..];

var A, Temp: [AllInds] real;

A[LastRow] = 1.0;

do {
  forall (i,j) in D do
    Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;

  const delta = max reduce abs(A[D] - Temp[D]);
  A[D] = Temp[D];
} while delta > epsilon;

writeln(A);
```

Wrap up

- assign Temp back to A for next iteration
- see whether we terminate using normal do..while loop
- print out final array once we're done

JACOBI ITERATION IN CHAPEL (NAMED, TUPLE-INDEXED VARIANT)

```
config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1},
      D = AllInds[1..n, 1..n],
      LastRow = AllInds[n+1..n+1, ..];

var A, Temp: [AllInds] real;

const north = (-1,0), south = (1,0), east = (0,1), west = (0,-1);

A[LastRow] = 1.0;

do {
  forall ij in D do
    Temp[ij] = (A[ij+north] + A[ij+south] + A[ij+east] + A[ij+west]) / 4;

    const delta = max reduce abs(A[D] - Temp[D]);
    A[D] = Temp[D];
  } while delta > epsilon;

writeln(A);
```

JACOBI ITERATION IN CHAPEL (SLICE-BASED VARIANT)

```
config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1},
      D = AllInds[1..n, 1..n],
      LastRow = AllInds[n+1..n+1, ..];

var A, Temp: [AllInds] real;

const north = (-1,0), south = (1,0), east = (0,1), west = (0,-1);

A[LastRow] = 1.0;

do {
  Temp[D] = (A[D.translate(north)] + A[D.translate(south)] +
            A[D.translate(east)] + A[D.translate(west)]) / 4;

  const delta = max reduce abs(A[D] - Temp[D]);
  A[D] = Temp[D];
} while delta > epsilon;

writeln(A);
```

JACOBI ITERATION IN CHAPEL (BACK TO THE SIMPLE, LOCAL VERSION)

```
config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1},
      D = AllInds[1..n, 1..n],
      LastRow = AllInds[n+1..n+1, ..];

var A, Temp: [AllInds] real;

A[LastRow] = 1.0;

do {
  forall (i,j) in D do
    Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;

  const delta = max reduce abs(A[D] - Temp[D]);
  A[D] = Temp[D];
} while delta > epsilon;

writeln(A);
```

JACOBI ITERATION IN CHAPEL (DISTRIBUTED VERSION)

```
use BlockDist;

config const n = 6, epsilon = 0.01;

const AllInds = {0..n+1, 0..n+1} dmapped Block({1..n, 1..n}),
      D = AllInds[1..n, 1..n],
      LastRow = AllInds[n+1..n+1, ..];

var A, Temp: [AllInds] real;

A[LastRow] = 1.0;

do {
  forall (i,j) in D do
    Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;

  const delta = max reduce abs(A[D] - Temp[D]);
  A[D] = Temp[D];
} while delta > epsilon;

writeln(A);
```


JACOBI ITERATION IN CHAPEL (DISTRIBUTED VERSION)

```
use BlockDist;  
  
config const n = 6, epsilon = 0.01;  
  
const AllInds = {0..n+1, 0..n+1} dmapped Block({1..n, 1..n}),  
      D = AllInds[1..n, 1..n],  
      LastRow = AllInds[n+1..n+1, ..];  
  
var A, Temp: [AllInds] real;
```

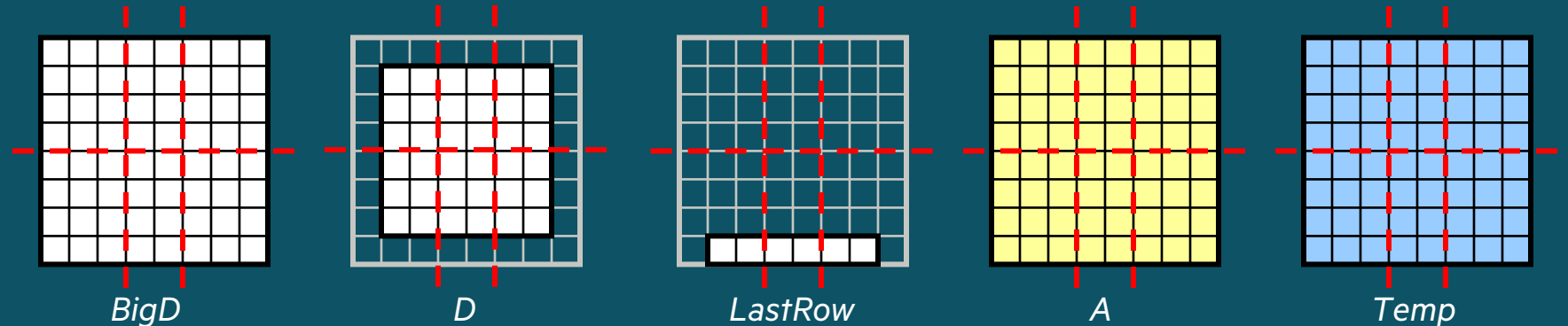
```
A[LastRow] = 1.0;
```

```
do {  
  forall (i,j) in D do  
    Temp[i,j] = (A[i-1,  
  
  const delta = max red  
  A[D] = Temp[D];  
} while delta > epsilon
```

```
writeln(A);
```

With these two changes, we distribute our domains, and therefore our computation

- Domain slices inherit parent domain's distribution



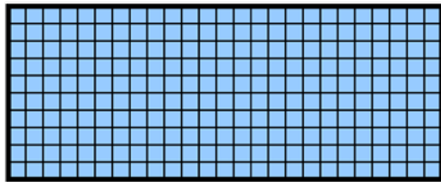
A FINAL NOTE ON THESE JACOBI EXAMPLES

- The previous slides were developed primarily to demonstrate data parallel features in Chapel
 - not necessarily to suggest “this is the best way to do Jacobi in Chapel”
 - specifically, we haven’t done any benchmarking or tuning of Jacobi as it hasn’t been of deep interest to our users
- If one wanted to do Jacobi in Chapel, there are a few other approaches to consider:
 - there’s a 'Stencil' distribution that is similar to 'Block' yet with a notion of ghost cells for caching neighbor values
 - and if one were to do a comparison, it’d be good to compare with a more manual SPMD version in Chapel as well

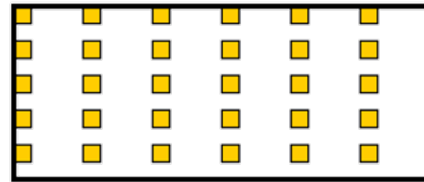


OTHER DATA PARALLEL FEATURES

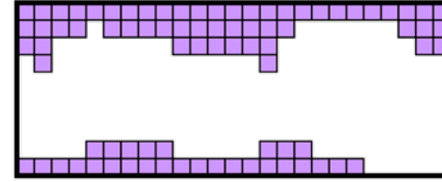
- **Scans:** parallel prefix operations
- **User-defined Parallel Iterators, Reduce/Scan Operations**
- **Several Domain/Array Types:**



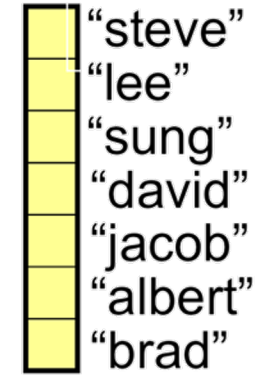
dense



strided



sparse



associative



SUMMARY OF THIS SECTION

- Chapel supports a rich set of language features
 - a modern, productive set of base language features
 - “low-level” features for creating tasks and placing them on a system
 - a global namespace for referring to data lexically, whether local or remote
 - high-level data-parallel features such as forall loops and promotion
 - a rich set of domains and arrays, including global-view distributed arrays

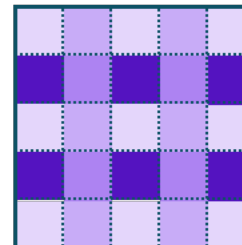
```
fib.chpl
config const n = 10;

for (i,f) in zip(0..<n, fib(n)) do
  writeln("fib #", i, " is ", f);

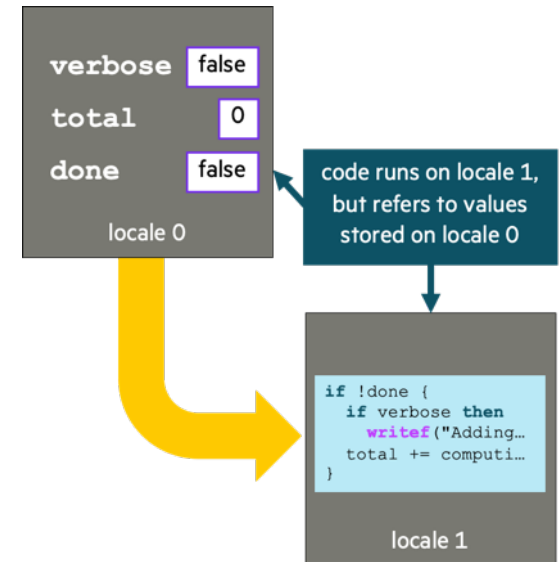
iter fib(x) {
  var current = 0,
      next = 1;

  for i in 1..x {
    yield current;
    current += next;
    current <=> next;
  }
}
```

```
prompt> chpl fib.chpl
prompt> ./fib --n=1000
fib #0 is 0
fib #1 is 1
fib #2 is 1
fib #3 is 2
fib #4 is 3
fib #5 is 5
fib #6 is 8
fib #7 is 13
fib #8 is 21
fib #9 is 34
fib #10 is 55
fib #11 is 89
fib #12 is 144
fib #13 is 233
fib #14 is 377
...
```



1.1	1.3	1.5	1.5	1.9
2.1	2.3	2.5	2.7	2.9
3.1	3.3	3.5	3.7	3.9
4.1	4.3	4.5	4.7	4.9
5.1	5.3	5.5	5.7	5.9



CHAPEL SOUNDS GREAT...THERE MUST BE A CATCH?

- We think Chapel is great, yet at times, it can admittedly also be frustrating
 - Compile times can feel sluggish (but we're currently working on improving them)
 - Error messages can be confusing or poor (ditto)
 - Chapel code doesn't support GPUs very well yet (ditto)
 - Sometimes reasonable code performs poorly (ditto)
 - Tools are lacking (not receiving much attention at present)
- Essentially, Chapel is a continually improving work-in-progress
 - Depending on your needs and personality, it may be perfect for you today, or it could make sense to wait
 - We have a reputation for being very responsive to users' questions and needs
- Another catch is that any language, however great, must overcome social challenges to become adopted
 - The HPC community is particularly skeptical of new languages
 - in part due to being performance- and HW-centric; in part due to having been burnt by past language attempts
- All that said, we think that the number of HPC-focused programming languages should be > 0
 - And that Chapel is as strong a contender as any



**LIVE DEMO?
(TIME AND INTEREST PERMITTING)**



CHAPEL ON OOKAMI



CHAPEL ON OOKAMI

- Chapel runs on Ookami
 - Our December release (1.25.1) is pre-installed as a module for users' convenience
 - We'll talk more about this in the hands-on section
- We have performed some baseline performance measurements
...but let's cover some disclaimers first



CHAPEL ON OOKAMI: IMPORTANT DISCLAIMERS

- We have not put *any* effort into tuning Chapel for A64FX processors
 - Our team spends a lot of effort tuning and optimizing Chapel for Cray and InfiniBand networks
 - And, to a lesser extent, optimizing for recent Intel and AMD processor designs
 - To date, A64FX has not been a priority for us
- We also haven't focused much on optimizing vectorization in Chapel
 - Our approach is to expose opportunities for vectorization to the back-end compiler, relying on it
 - recently, we have been focused on code generation for GPUs which is related, but different
- A64FX is, in many respects, the opposite of what we've been most focused on in recent years:
 - **Our recent focus:** massive data sets on systems with lots of memory and bandwidth
 - **A64FX:** memory capacity limited, vectorization-focused



BASELINE OOKAMI PERFORMANCE COMPARISONS

16-node Chapel results:

Benchmark	Apollo-CL	Ookami	Ratio
Stream Triad	2579 GB/s	7808 GB/s	3.03x
PRK Stencil	1335 GFlops/s	949 GFlops/s	0.71x
ISx	2.55 s	6.86 s	0.37x
Bale IndexGather	4.8 GB/s/node	0.4 GB/s/node	0.08x
HPCC RA (RMA)	0.0016 GUPS	0.0009 GUPS	0.56x
HPCC RA (AM)	0.0084 GUPS	0.0024 GUPS	0.29x

Ookami's HBM greatly benefits highly localized computations

System Characteristics:

System	Network	Cores per node (locale)	Processor Type
Apollo-CL	HDR IB	48	Cascade Lake
Ookami	HDR IB	48	A64FX



(see the disclaimers on the preceding slide before drawing any conclusions from these results)

BASELINE OOKAMI PERFORMANCE COMPARISONS

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Possible explanations for the poor results for more complex benchmarks:

- Chapel and its communication optimizations may require more powerful scalar cores than those on A64FX
- Chapel's heuristics for NUMA affinity may be less effective on A64FX than on CascadeLake
- Lack of vectorization / CPU specialization may hurt Chapel more on A64FX than on CascadeLake
- Chapel's tasking library (Qthreads) may not perform as well on A64FX as on CascadeLake

POTENTIAL OOKAMI IMPROVEMENTS: MULTIPLE PROCESSES PER NODE

- Traditionally, Chapel runs a single process per locale per compute node
 - Parallelism is typically implemented via user-level tasks
 - executed using worker threads that are pinned 1:1 to the compute node’s cores
 - NUMA affinity is dealt with heuristically by Chapel’s implementation
 - not perfect, but has typically worked “well enough” in practice
 - This approach has had various benefits for us, including:
 - simple execution model for users
 - single communication mechanism for cross-locale accesses
 - good surface-to-volume properties, particularly as core counts have increased significantly
- For various reasons, we have discussed enabling a slightly coarser execution model
 - e.g., running using a process/locale per...
 - ...NUMA domain / CMG?
 - ...NIC?
 - ...GPU?
- If NUMA effects are hurting Chapel on A64FX more than conventional processors, this could help



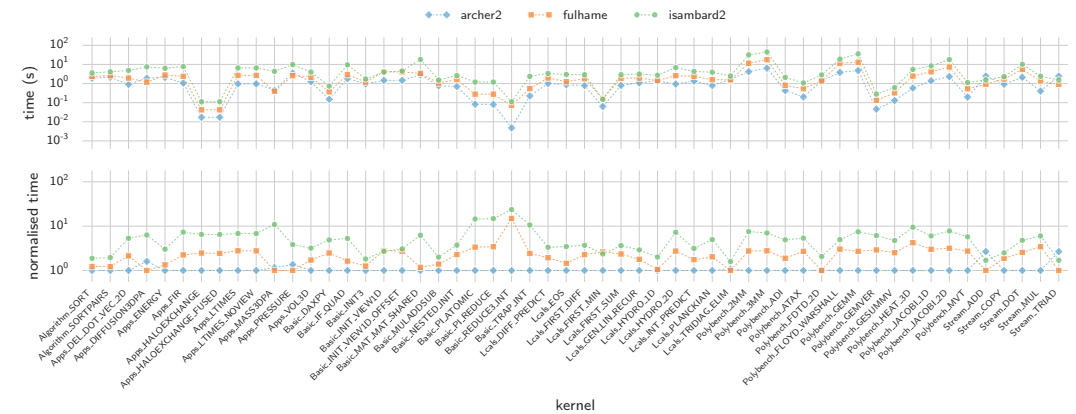
POTENTIAL OOKAMI IMPROVEMENTS: BETTER VECTORIZATION

- As mentioned earlier, vectorization optimizations have not been a big focus for our group to date
- Three promising directions:
 1. Simon Moll et al.'s LLVM Region Vectorizer has demonstrated benefits for Chapel via outer-loop vectorization
 - we'd like to explore this more and potentially enable it by default in Chapel
 2. ARM has been upstreaming SVE contributions to LLVM that could also improve our performance on A64FX
 - thanks to Tony Curtis for bringing this to our attention
 3. colleagues at EPCC have recently started a study of vectorization on ARM-based HPC systems
 - (see next slide)



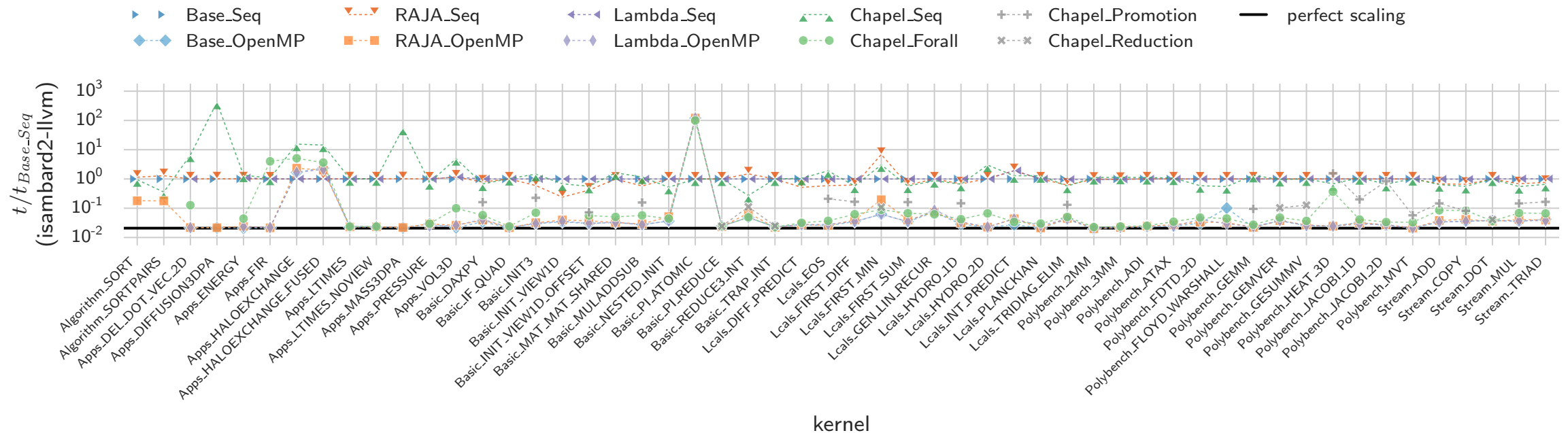
POTENTIAL MITIGATION STRATEGIES: BETTER VECTORIZATION

- Michele Weiland & Ricardo Jesus (EPCC): studying how well languages target ARM-based HPC systems
 - Using [RAJAPerf](#) as the basis for their study
 - Have created a Chapel port using a few different computational styles
 - Running on ARCHER2 (AMD EPYC 7742), Fulhame (Marvell ThunderX2), and Isambard 2 (Fujitsu A64FX)
 - Starting to generate preliminary results:



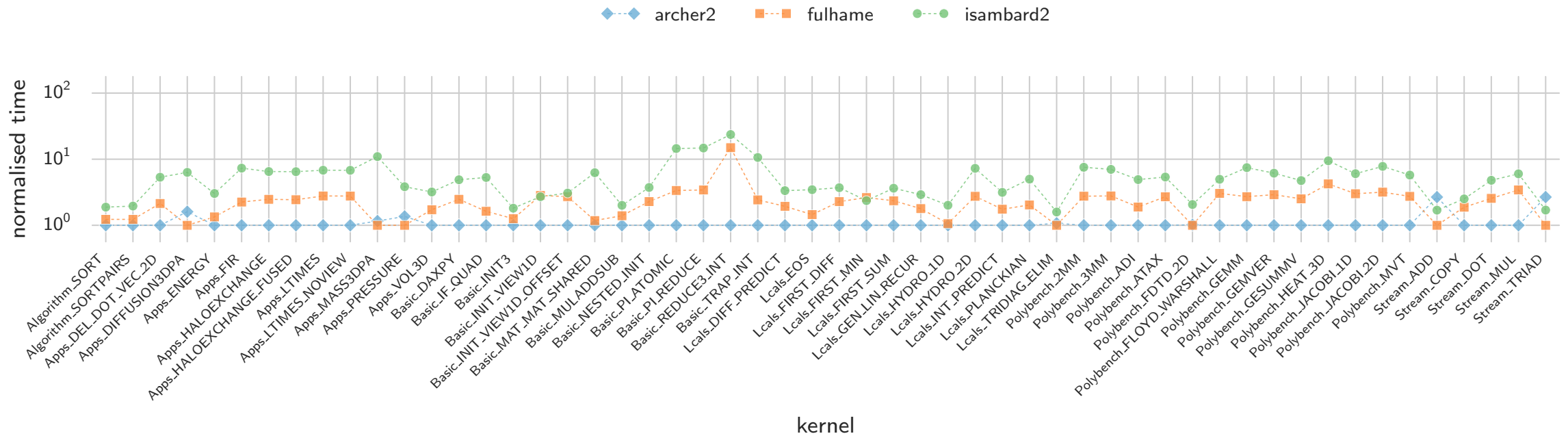
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PRK Stencil	1335 GFlops/s	949 GFlops/s	0.71x	Ookami-SHMEM
ISx	2.55 s	6.86 s	0.37x	5.76 s
Bale IndexGather	4.8 GB/s/node	0.4 GB/s/node	0.08x	0.47 GB/s/node
HPCC RA (RMA)	0.0016 GUPS	0.0009 GUPS	0.56x	
HPCC RA (AM)	0.0084 GUPS	0.0024 GUPS	0.29x	

SHMEM reference versions perform similarly to Chapel

- these are run with a process per core, so don't suffer NUMA effects
- may suggest that scalar processor / communication overheads are the likely big difference



(see the disclaimers on the preceding slide before drawing any conclusions from these results)

CHAPEL ON OOKAMI PERFORMANCE SUMMARY

- Stating for re-emphasis: We've invested no effort to date in making Chapel perform well on Ookami
- That said, we see several potential avenues for improvements:
 - better vectorization / specialization for A64FX
 - improve our understanding of how Chapel's communication code paths behave on A64FX
 - running a process/locale per CMG ; better NUMA heuristics
 - studying and tuning the performance of the Qthreads tasking library on A64FX
 - ...something else we haven't learned yet due to lack of study?



SUMMARY & RESOURCES



SUMMARY

Chapel cleanly supports...

- ...expression of parallelism and locality
- ...a diverse set of parallel features, at various levels
- ...specifying how to map computations to the system

Chapel is powerful:

- it supports succinct, straightforward code
- it can result in performance that competes with or beats C+MPI[+OpenMP]

Chapel is being used for productive parallel applications at scale

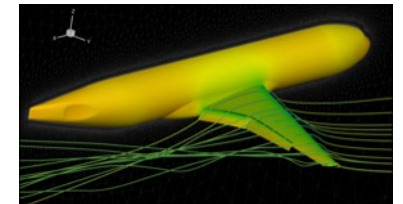
- recent users have reaped its benefits in 10k–48k-line applications

Chapel is available and working on Ookami

- further study is required to understand opportunities for improved performance

Why Consider New Languages at all?

Syntax	<ul style="list-style-type: none">• High level, elegant syntax• Improve programmer productivity
Semantics	<ul style="list-style-type: none">• Static analysis can help with correctness• We need a compiler (front-end)
Performance	<ul style="list-style-type: none">• If optimizations are needed to get performance• We need a compiler (back-end)
Algorithms	<ul style="list-style-type: none">• Language defines what is easy and hard• Influences algorithmic thinking



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AGAIN, WE ARE HIRING

Chapel Development Team at HPE



see: <https://chapel-lang.org/jobs.html>



CHAPEL RESOURCES

Chapel homepage: <https://chapel-lang.org>


- (points to all other resources)

Social Media:

- Twitter: [@ChapelLanguage](https://twitter.com/ChapelLanguage)
- Facebook: [@ChapelLanguage](https://www.facebook.com/ChapelLanguage)
- YouTube: <http://www.youtube.com/c/ChapelParallelProgrammingLanguage>

Community Discussion / Support:

- Discourse: <https://chapel.discourse.group/>
- Gitter: <https://gitter.im/chapel-lang/chapel>
- Stack Overflow: <https://stackoverflow.com/questions/tagged/chapel>
- GitHub Issues: <https://github.com/chapel-lang/chapel/issues>



The Chapel Parallel Programming Language

What is Chapel?

Chapel is a programming language designed for productive parallel computing at scale.

Why Chapel?

Because it simplifies parallel programming through elegant support for:

- **distributed arrays** that can leverage thousands of nodes' memories and cores
- a **global namespace** supporting direct access to local or remote variables
- **data parallelism** to trivially use the cores of a laptop, cluster, or supercomputer
- **task parallelism** to create concurrency within a node or across the system

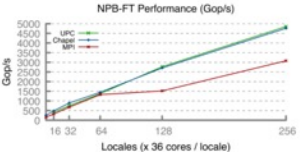
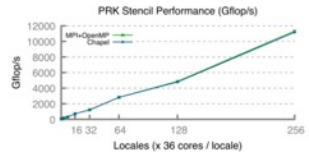
Chapel Characteristics

- **productive**: code tends to be similarly readable/writable as Python
- **scalable**: runs on laptops, clusters, the cloud, and HPC systems
- **fast**: performance *competes with or beats* C/C++ & MPI & OpenMP
- **portable**: compiles and runs in virtually any *nix environment
- **open-source**: hosted on GitHub, permissively licensed

New to Chapel?

As an introduction to Chapel, you may want to...

- watch an [overview talk](#) or browse its [slides](#)
- read a [blog-length](#) or [chapter-length](#) introduction to Chapel
- learn about [projects powered by Chapel](#)
- check out [performance highlights](#) like these:



The PRK Stencil Performance graph shows Chapel (green line) significantly outperforming MPI+OpenMP (red line) as the number of locales increases from 16 to 256. The NPB-FT Performance graph shows Chapel (green line) also outperforming MPI (red line) in a similar trend.

- browse [sample programs](#) or learn how to write distributed programs like this one:

```
use CyclicDist;           // use the Cyclic distribution library
config const n = 100;     // use --n=<val> when executing to override this default

forall i in {1..n} dmapped Cyclic(startIdx=1) do
  writeln("Hello from iteration ", i, " of ", n, " running on node ", here.id);
```

SUGGESTED READING / VIEWING

Chapel Overviews / History (in chronological order):

- [Chapel](#) chapter from [Programming Models for Parallel Computing](#), MIT Press, edited by Pavan Balaji, November 2015
- [Chapel Comes of Age: Making Scalable Programming Productive](#), Chamberlain et al., CUG 2018, May 2018
- Proceedings of the [8th Annual Chapel Implementers and Users Workshop](#) (CHI UW 2021), June 2021
- [Chapel Release Notes](#) — current version 1.25, October 2021

Arkouda:

- Bill Reus’s CHI UW 2020 keynote talk: <https://chapel-lang.org/CHI UW 2020.html#keynote>
- Arkouda GitHub repo and pointers to other resources: <https://github.com/Bears-R-Us/arkouda>

CHAMPS:

- Eric Laurendeau’s CHI UW 2021 keynote talk: <https://chapel-lang.org/CHI UW 2021.html#keynote>
 - two of his students also gave presentations at CHI UW 2021, also available from the URL above
- Another paper/presentation by his students at <https://chapel-lang.org/papers.html> (search “Laurendeau”)



**HANDS-ON
(TIME AND INTEREST PERMITTING)**



USING CHAPEL ON OOKAMI

- Chapel is pre-installed on Ookami, thanks to Eva Siegmann and Tony Curtis
 - Installed at `/lustre/software/chapel/apollo/chapel-1.25.1`
 - Available via normal module commands:

```
prompt> module load chapel
```

- Sample programs available:

```
prompt> ls $CHPL_HOME/examples/*.chpl
hello.chpl                hello4-datapar-dist.chpl
hello2-module.chpl        hello5-taskpar.chpl
hello3-datapar.chpl       hello6-taskpar-dist.chpl
```

(see also the ‘primers/’ and ‘benchmarks/’ subdirectories)

- Compile and run as shown in previous examples:

```
prompt> chpl $CHPL_HOME/examples/hello6-taskpar-dist.chpl
prompt> ./hello6-taskpar-dist -nl 4
```

INSTALLING AND USING CHAPEL ON YOUR OWN SYSTEM

- Often, getting started with Chapel on a supercomputer can be annoying
 - Environment not as set up to your liking as your primary machine
 - Shared resource, queueing times, etc.
- You're welcome to install Chapel on your laptop or favorite system if you're able to
 - **Mac homebrew** (Catalina or later) / **Linuxbrew**: 'brew install chapel' (supports single-locale runs only)
 - **Mac / Linux / *nix**: Install and build from source, see <https://chapel-lang.org/download.html>
 - **Windows**: Use Linux bash shell / Windows Subsystem for Linux and see previous line
 - **Docker**: See <https://chapel-lang.org/install-docker.html>
- Developing a Chapel program on a laptop and then running it on a supercomputer is a common practice
 - And Chapel's global view tends to make it easy:
 - almost always runs correctly
 - typically not too difficult to get using multiple locales, particularly for data-parallel codes
 - optimizing it can take more effort...



“WHAT SHOULD I DO DURING THE HANDS-ON SESSION?”

- You're welcome to do whatever appeals to you, but here are some possibilities:

- Try compiling, running, modifying examples from this talk

- I've made most of them available on Ookami at:

- `/lustre/projects/global/Chapel/ookami-webinar/slideExamples/`

- Try one of the hands-on exercises

- There are four exercises prepared, two that are simpler and two that are more involved (see next slide)

- Instructions and starting files are also on Ookami:

- `/lustre/projects/global/Chapel/ookami-webinar/handsOn/`

- Try coding up a computation or parallel pattern of interest to you

- Note that you will probably want to create your own local, writeable copy of the materials above, e.g.:

```
prompt> cp -r /lustre/projects/global/Chapel/ookami-webinar .
```

```
prompt> chmod -R u+w ookami-webinar
```



PREPARED HANDS-ON EXERCISES

- **Advent of Code 2021 Day 1:** given an array of numbers, compute some simple statistics on it
 - This is a trivial computation that you'd have no trouble doing in a language you're familiar with
 - Goal is to use it to get familiar with Chapel
 - Opportunities for array operations, data parallelism, reductions
- **Advent of Code 2021 Day 4:** simulate a bingo game with an octopus
 - This is slightly more involved and interesting, but still straightforward
 - More opportunities for array operations, data parallelism, 2D arrays
- **Ray Tracer:** given a ray tracing framework, fill in some missing details to make it work
 - Exercises 2D arrays, data parallelism
- **Bounded Buffer:** use Chapel's sync and atomic variables
 - Exercises task parallelism and synchronization—relies on sync/atomic variables, not really covered today

(Note that none of these runs are large enough to *require* multiple locales, though most are amenable to them)



NEED HELP?

- We'll be handling Q&A today both live and via #chapel-webinar on the IACS Slack channel
 - Members of the Chapel team besides myself will be on Slack to answer questions, screen share, etc.
- After today, help is available via:
 - **Stack Overflow:** for questions that will likely be valuable to others (tag your question with 'chapel')
 - **Discourse:** our community web forum / mailing list technology
 - **Gitter:** our community real-time chat technology
 - **GitHub issues:** for filing bug reports, feature requests, etc.



GENERAL TIPS WHEN GETTING STARTED WITH CHAPEL (ALSO IN README)

- Online **documentation** is here: <https://chapel-lang.org/docs/>
 - The primers can be particularly valuable for learning a concept: <https://chapel-lang.org/docs/primers/index.html>
 - These are also available from a Chapel release in ‘\$CHPL_HOME/examples/primers/’
 - or ‘\$CHPL_HOME/test/release/examples/primers/’ if you clone from GitHub
- When debugging, **almost anything in Chapel can be printed out** with ‘writeln(expr1, expr2, expr3);’
 - Types can be printed after being cast to strings, e.g. ‘writeln(“Type of “, expr, “ is “, expr.type:string);’
 - A quick way to print a bunch of values out clearly is to print a tuple made up of them ‘writeln((x, y, z));’
- Once your code is correct, before doing any performance timings, be sure to re-compile with ‘**--fast**’
 - Turns on optimizations, turns off safety checks, slows down compilation, speeds up execution significantly
 - Then, when you go back to making modifications, be sure to stop using ‘--fast’ in order to turn checks back on
- For vim / emacs users, **syntax highlighters** are in \$CHPL_HOME/highlight
 - Imperfect, but typically better than nothing
 - Emacs MELPA users may want to use the chapel-mode available there (better in many ways, weird in others)



THANK YOU

<https://chapel-lang.org>
@ChapelLanguage

