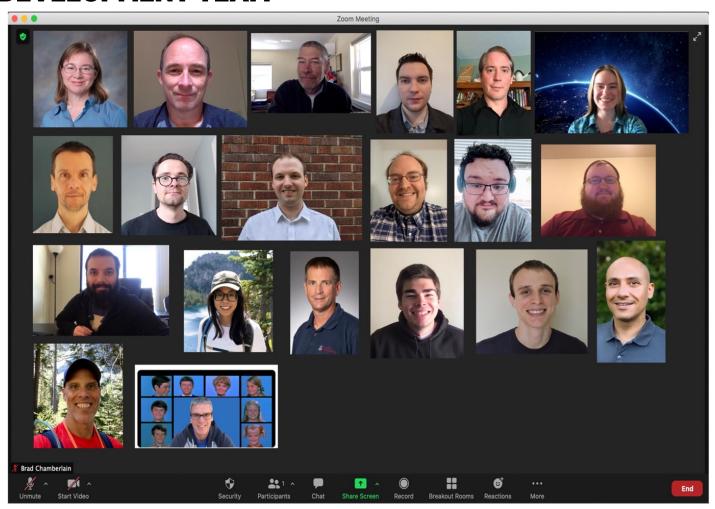
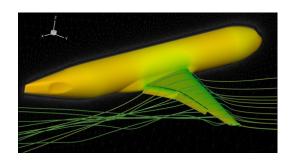


## **CHAPEL DEVELOPMENT TEAM**



## **HIGH PERFORMANCE COMPUTING**

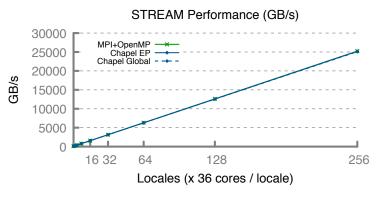




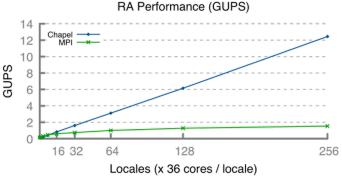
- Simulations
- Analysis of massive datasets
- Ever-changing machines
- Challenging for programmer productivity



## CHAPEL BENCHMARKS TEND TO BE CONCISE, CLEAR, AND PERFORMANT







## TAKEAWAY: CAN SEPARATE PERFORMANCE CONCERNS WITH CHAPEL

# STREAM Triad: $\vec{A} = \vec{B} + \alpha \vec{C}$

```
#include <hpcc.h>
#include <omp.h>
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 );
  // deleted error malloc error checking code
  #pragma omp parallel for
  for (j=0; j<VectorSize; j++) {</pre>
   b[j] = 2.0;
    c[j] = 1.0;
  scalar = 3.0;
  #pragma omp parallel for
  for (j=0; j<VectorSize; j++)</pre>
   a[j] = b[j] + scalar*c[j];
  // deleted deallocation code
  return 0;
```

**Goal**: Let programmers **control** performance concerns in a **separate** stack of abstractions.

**Talk**: Example abstractions in Chapel, including for GPUs

## **HPC PROGRAMMERS WANT CONTROL**

## STREAM Triad: $\vec{A} = \vec{B} + \alpha \vec{C}$

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```

HPC programmers obtain control by coding at a lowlevel of detail

Let's provide separate **What** and **How** constructs and abstractions

**Goal**: Let programmers **control** performance concerns in a **separate** stack of abstractions.

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  #pragma omp parallel for
  #pragma omp parallel for
  // deleted deallocation code
  return 0;
```

## What

• STREAM Triad computation

# STREAM Triad: $\vec{A} = \vec{B} + \alpha \vec{C}$

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  return 0;
```

- What
  - STREAM Triad computation
- How
  - Data organization across nodes in a parallel machine

# STREAM Triad: $\vec{A} = \vec{B} + \alpha \vec{C}$

```
#include <hpcc.h>
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```

- What
  - STREAM Triad computation
- How
  - Data organization across nodes in a parallel machine
  - Schedule: process per node (MPI, message passing interface)

# STREAM Triad: $\vec{A} = \vec{B} + \alpha \vec{C}$

#### What

• STREAM Triad computation

#### • How

- Data organization across nodes in a parallel machine
- Schedule: process per node (MPI, message passing interface)
- Schedule: threads per process (OpenMP)

## JUST HIDE DETAILS IN WHAT CODE FUNCTIONS

# STREAM Triad: $\vec{A} = \vec{B} + \alpha \vec{C}$

```
#include <hpcc.h>
#include <omp.h>
static int VectorSize;
static double *a, *b, *c;
int HPCC_StarStream(HPCC_Params *params) {
 return errCount;
int HPCC Stream(HPCC Params *params, int doIO) {
register int i:
  a = HPCC_XMALLOC( double, VectorSize )
  b = HPCC_XMALLOC( double, VectorSize )
  c = HPCC XMALLOC( double, VectorSize )
  // deleted error malloc error checking co
 #pragma omp parallel for
#pragma omp parallel for
  // deleted deallocation code
 return 0;
```

```
Vector*
STREAM_Triad(Vector *B, double alpha,
    Vector* C) {
    ...
    ...
}
Vector* A = STREAM_Triad(B, 2.0, C);
```

**Doesn't work!** Composition? New Machine?

## JUST RAISE THE LEVEL OF ABSTRACTION



If we just raise the abstraction level in the **What** stack high-enough, the compiler or library can effectively map computations to a variety of hardware.

**Problem:** Programmer loses control

**Need**: Programmer control at multiple levels of a

How stack, multiresolution

## **CHAPEL: MULTIRESOLUTION PROGRAMMING EXAMPLE**

- Domain mapping construct "dmapped ..." semantics like HPF
- However, "Block" is implemented with userfacing constructs
- Users can write their own domain mappings
- Provides multiresolution programming

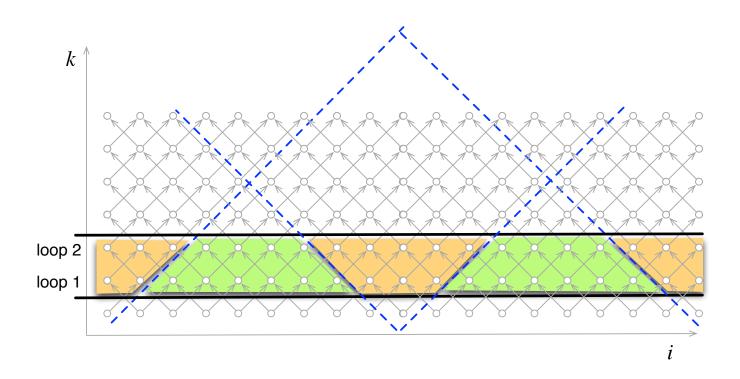
## CHAPEL ITERATORS PROVIDE MULTIRESOLUTION SCHEDULE CONTROL

- Programmer can see details of default iterator
- Programmer can write own domains mappings and iterators

```
// Default iterator for a Block domain
iter these ( ... ) {

coforall loc in Domain.targetLocales {
  on loc {
    const numTasks = computeNumTasks();
    const myInds = Domain.localSubdomain(loc);
    coforall tid in 0..<numTasks {
    const myChunk = chunk(myInds, tid, numTasks);
    for i in myChunk do
       yield i;
    }
}</pre>
```

## **MORE COMPLEX SCHEDULE: DIAMOND SLAB TILING**



## **MAKING SCHEDULES AVAILABLE IN LIBRARIES**

## Diamond slab tiling written in C

## Diamond slab tiling made available as a Chapel iterator

```
int Li=0, Ui=N, Lj=0, Uj=N;
    for(int ts=0; ts<T; ts+=subset s)
    for (int c0 = -2; c0<=0; c0+=1)
       for (int c1 = 0; c1 <= (Uj+tau-3)/(tau-3); c1+=1)
        for (int x = (-Ui-tau+2)/(tau-3); x <= 0; x += 1) {
          int c2 = x-c1; //skew
          // loops for time steps within a slab
          // (slices within slabs)
          for (int c3 = 1; c3 <= subset s; c3 += 1)
            for (int c4 = max(max(max(-tau * c1 - tau *
c2 + 2 * c3 - (2*tau-2), -Uj - tau * c2 + c3 - (tau-2)),
tau * c0 - tau * c1 - tau * c2 - c3), Li); c4 <=
min(min(min(tau * c0 - tau * c1 - tau * c2 - c3 + (tau-
1), -\tan * c1 - \tan * c2 + 2 * c3, -L_1 - \tan * c2 + c3,
Ui - 1); c4 += 1)
             for (int c5 = max(max(tau * c1 - c3, Lj), -
tau * c2 + c3 - c4 - (tau-1)); c5 <= min(min(Uj - 1, -tau))
* c2 + c3 - c4), tau * c1 - c3 + (tau-1)); c5 += 1)
```

```
We want to transform our original schedule:
for t in timeRange do
    forall (x,y) in spaceDomain do
        computation(t, x, y);
into a faster schedule:
  forall (t,x,y)
  in diamondTileIterator(...) do
        computation(t, x, y);
```

Ian J. Bertolacci, Catherine Olschanowsky, Ben Harshbarger, Bradford L. Chamberlain, David G. Wonnacott, and Michelle Mills Strout. "Parameterized Diamond Tiling for Stencil Computations with Chapel Parallel Iterators." In the Proceedings of the 29th International Conference on Supercomputing (ICS), June 2015.



## **PLANS FOR GPU SUPPORT**

Vision

## **Memory/Locality Management**

- Chapel's locale model concept supports describing a compute node with GPU naturally
  - The execution and memory allocations can be moved to GPU sublocales
- Arrays can be declared inside 'on' statements to allocate them on GPU memory
- Or distributed arrays that target GPU sublocales can be created

# Done in 1.25

#### **Execution**

- Chapel's order-independent loops (i.e., 'forall' and 'foreach') can be transformed into GPU kernels
  - If such a loop is encountered while executing on a GPU sublocale, the corresponding kernel is launched
  - GPU code is generated for every call inside the loop body

### **Other Possible Multiresolution Features**

- Specifying the grid and block organization for the GPU version of the computation
- Queries about node architecture including number and kind of GPUs to guide iteration and data org

## **CHAPEL 1.25 EFFORT**

Putting the Pieces Together

#### User's loop

```
forall i in 1..n do arr[i] = i*mul;
```

#### The loop is replaced with:

#### **Generated GPU kernel looks like:**

```
pragma "codegen for GPU"
proc kernel(in startIdx, in endIdx,
           ref arrArg, in mulArg) {
 var blockIdxX = primitive('gpu blockIdx x')
 var blockDimX = primitive('gpu blockDim x')
 var threadIdxX = primitive('gpu threadIdx x')
 var t0 = blockIdxX * blockDimX
 var t1 = t.0 + threadIdxX
 var index = t1 + startIdx
 var chpl is oob = index > endIdx
 if (chpl is oob) { return; }
 var arrData = arrArg->data
 ref addrToChange = &arrData[index]
 var newVal = myIdx*mulArg
  *addrToChange = newVal
```

## **STATUS**

Stream

```
on here.getChild(1) {
  var a, b, c: [1..n] real;
  const alpha = 2.0;

b = 1.0;
  c = 2.0;

Promotion (e.g., 'b = 1.0') still executes on host

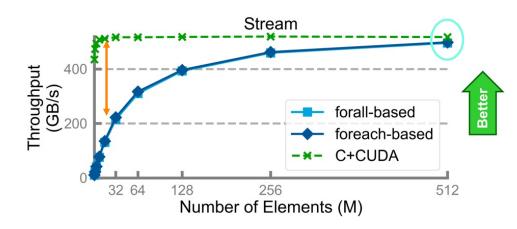
forall aElem, bElem, cElem in zip(a, b, c) do
  aElem = bElem + alpha * cElem;
//or
forall i in a.domain do
  a[i] = b[i] + alpha * c[i];
• Scalars are allocated on the function stack
• So, they are on host memory

These foralls will execute on GPU
```

• Arrays are allocated in unified memory

## STATUS OF GPU PERFORMANCE FOR CHAPEL

An Early Performance Study



At smaller vector sizes throughput is low

At larger vector sizes efficiency reaches 96%

## **Takeaways**

- No major performance-related issue in the prototype
- Gets close to 100% efficiency with large datasets
- 'foreach' is slightly faster than 'forall'

#### **Potential Sources of Overhead**

- I/O for loading the GPU kernel for each launch
- Unified memory vs device memory
- Kernel argument allocations

#### **Prospects**

- Generating single binary will remove the I/O cost
- Profile the remaining costs
- Implement other benchmarks

## **SUMMARY AND THANK YOU!**

michelle.strout@hpe.com

- **Goal**: Let programmers **control** performance concerns in a **separate** stack of abstractions
- Chapel's multiresolution control enables the separation of parallel performance concerns
  - Domain mappings
  - Iterators
- Multiresolution control in for GPU support in Chapel 1.25
- The Chapel team is hiring! (https://chapellang.org/jobs.html)

