

Hewlett Packard  
Enterprise

# ASYNCHRONOUS TASK-BASED AGGREGATED COMMUNICATION IN CHAPEL

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AMTE 2022  
August 23, 2022

# WHAT IS CHAPEL?

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**Chapel:** A modern parallel programming language

- portable & scalable
- open-source & collaborative



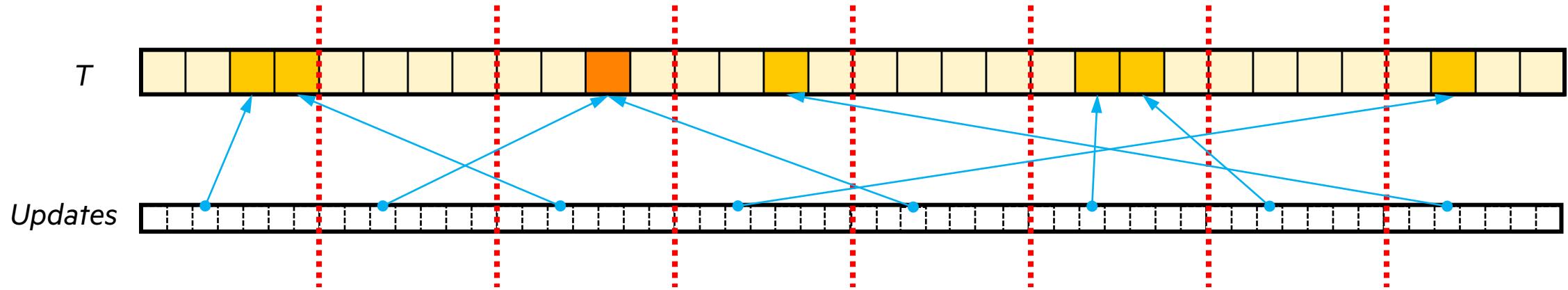
## Goals:

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



# HPCC RANDOM ACCESS (RA)

**Data Structure:** distributed table



**Computation:** in parallel, update random table elements with random values

**Declarations:** distributed table and index space of updates in Chapel:

```
var T: [newBlockDom(0..<tableSize)] atomic int;  
const Updates = newBlockDom(0..<numUpdates);
```



# 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 & (TABSIZ-1)] ^= Ran;
 * }
 */

MPI_Irecv(&LocalRecvBuffer, localBufferSize, tparams.dtype64,
          MPI_ANY_SOURCE, MPI_ANY_TAG, MPI_COMM_WORLD, &inreq);
while (i < 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 remaining updates in buckets*/
while (pendingUpdates > 0) {
    /*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) {
                /*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);
        }
    } while (have_done && NumberReceiving > 0);

    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;
    }
}

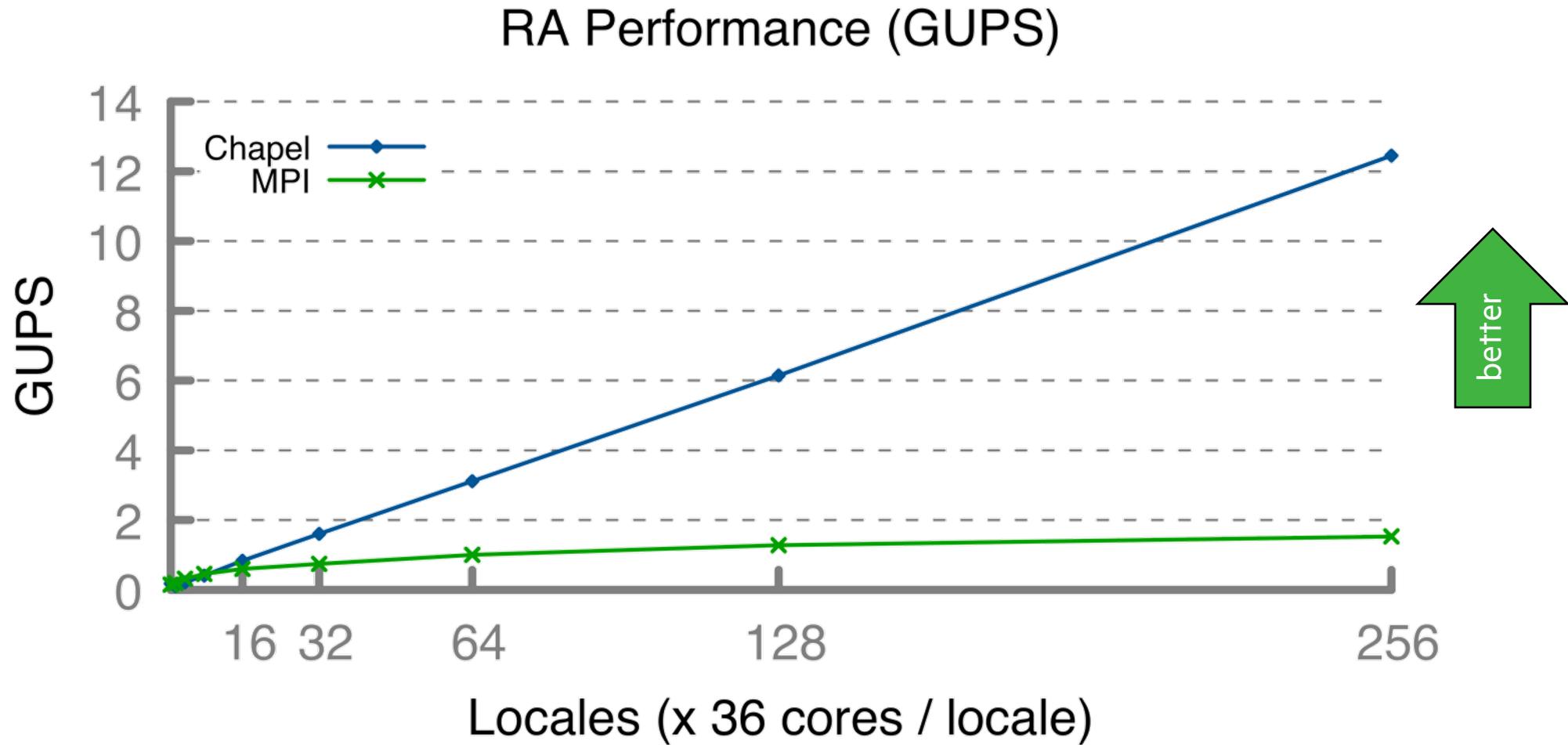
/*send our done messages*/
for (proc_count = 0 ; proc_count < tparams.NumProcs ; ++proc_count) {
    if (proc_count == tparams.MyProc) { tparams.finish_req[tparams.MyProc] =
                                         MPI_REQUEST_NULL; continue; }
    /* send garbage - who cares, no one will look at it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



# HPCC RA: CHAPEL VS. C+MPI



# HPCC RA IN CHAPEL: NAÏVE IMPLEMENTATION

```
/* 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 & (TABSIZ-1)] ^= Ran;  
 * }  
 */
```

```
MPI_Irecv(&LocalRecvBuffer, localBufferSize,  
          MPI_ANY_SOURCE, MPI_ANY_TAG, &inreq);  
while (i < 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++) {
```

```
coforall loc in Updates.targetLocales do  
    on loc do  
        coforall tid in 1..here.numPUs() do  
            for idx in myInds(loc, tid, ...) do  
                T[idx & indexMask].xor(idx);  
    }
```

```
    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++;
```

Chapel Kernel

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

Gets lowered roughly to...

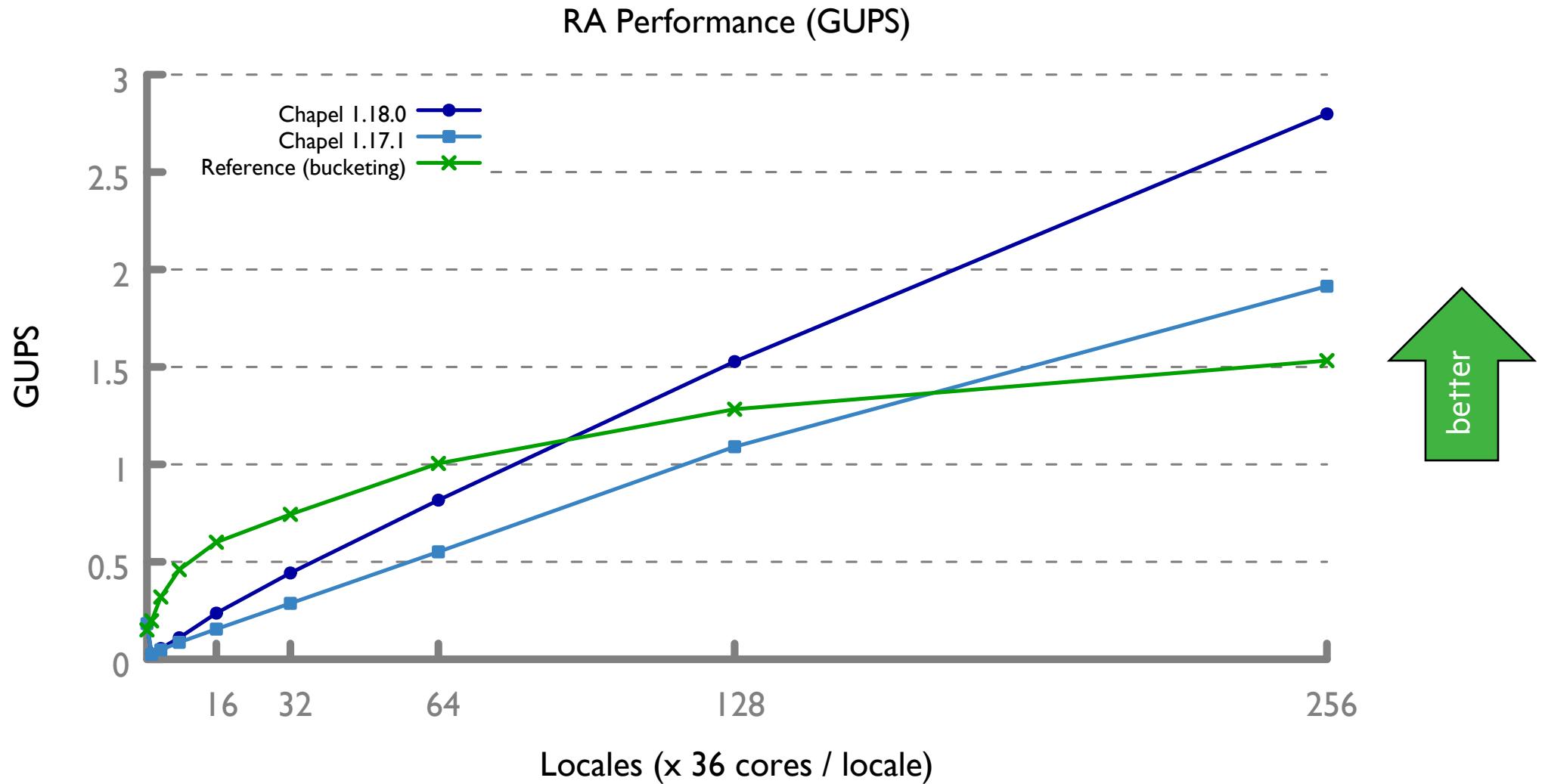
```
MPI_Test(&outreq, &have_done, MPI_STATUS_IGNORE);  
if (have_done) {  
    outreq = MPI_REQUEST_NULL;  
    MPI_Gatherv(&Ran, 1, MPI_UNSIGNED, &LocalRecvBuffer, localBufferSize,  
               MPI_COMM_WORLD, tparams.dtype64, (int)pe, &outreq);  
  
    proc_count = 0; proc_count < tparams.NumProcs ; ++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);
```

A concurrent loop over the compute nodes

A nested concurrent loop over each node's cores

A serial loop to compute each task's chunk of updates

# HPCC RA: NAÏVE CHAPEL VS. C+MPI (SEPTEMBER 2018)



# UNORDERED OPERATION OPTIMIZATION

```
/* 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 & (TABSIZ-1)] ^= Ran;  
 * }
```

```
MPI_Irecv(&LocalRecvBuffer, localBufSize,  
          MPI_ANY_SOURCE, MPI_A  
while (i < 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++) {  
                    /* do something with the data */  
                    pendingUpdates++;  
                }  
            }  
        }  
    } while (!have_done);  
}
```

```
coforall loc in Updates.targetLocales do  
    on loc do  
        coforall tid in 1..here.numPUs() do  
            for idx in myInds(loc, tid, ...) do  
                T[idx & indexMask].xor(idx);  
    }  
}
```

```
    minInd = (GlobalOffset / (tparams.MinLocalTableSize + 1));  
    else  
        WhichPe = ( (GlobalOffset - tparams.Remainder) /  
                    tparams.MinLocalTableSize );  
    if (WhichPe == tparams.NumLocales)  
        LocalOffset = 0;  
    HPCC_Table[Loc  
for idx in myInds(loc, tid, ...) do  
    T[idx & indexMask].unorderedXor(idx);  
    unorderedFence();
```

```
) else {  
    HPCC_InsertUpdate(Ran, WhichPe, Buckets);  
    pendingUpdates++;  
}  
i++;
```

Chapel Kernel

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

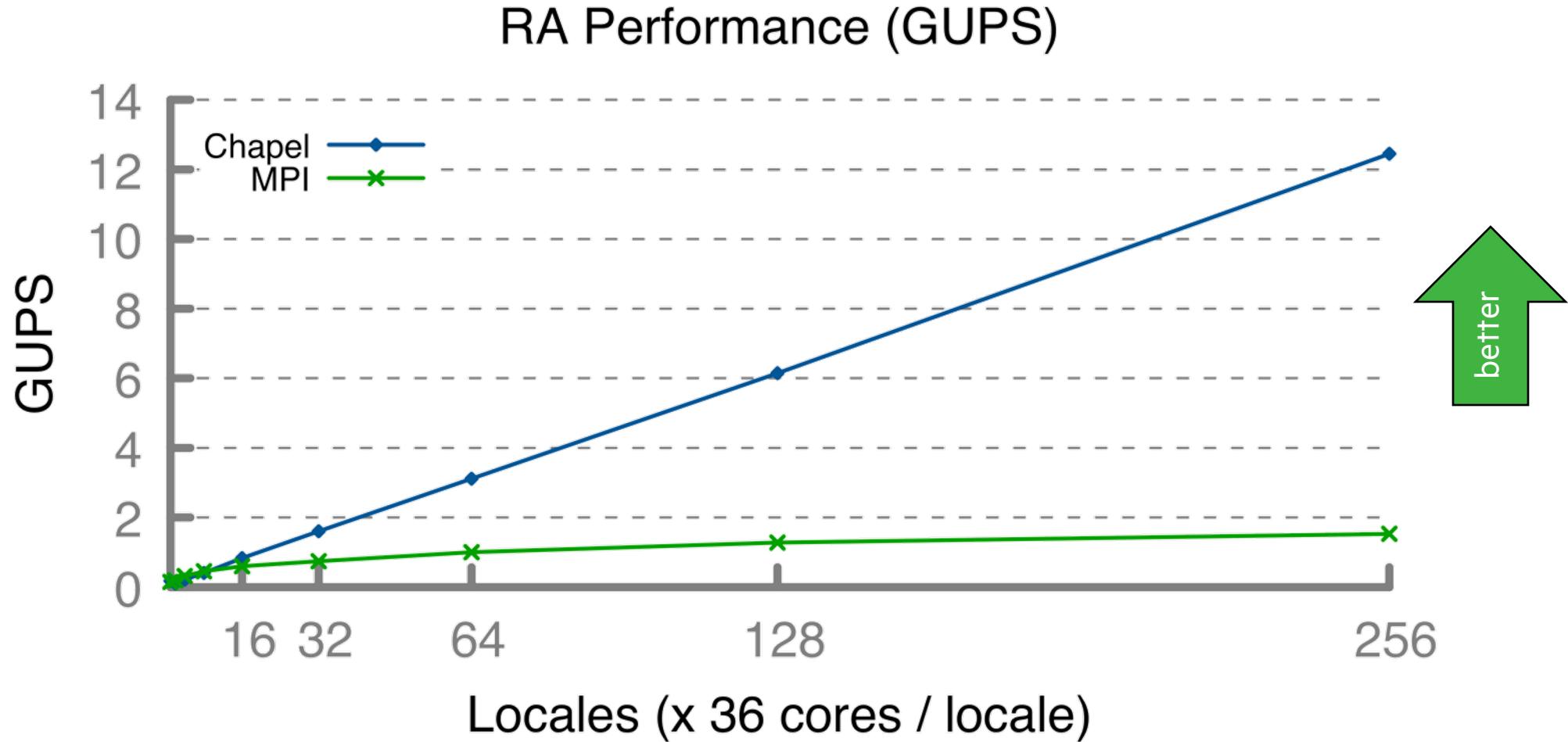
```
MPI_Test(&outreq, &have_done, MPI_STATUS_IGNORE);  
if (have_done) {  
    outreq = MPI_REQUEST_NULL;  
    /* do something with the data */  
    MPI_Irecv(&LocalRecvBuffer, localBufSize,  
              tparams.dtype64, (int)pe,  
              &outreq);
```

```
    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);  
    }  
    /* done 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 = LocalRecvBuff[bufferBase+j];  
            }  
        }  
        NumberReceiving--;  
    }  
    else {  
        MPI_Abort ( MPI_COMM_WORLD, -1 );  
    }  
}
```

But, for a parallel loop with no data dependencies,  
why perform these high-latency operations serially?

So, our compiler rewrites the inner loop  
to perform the ops asynchronously

# HPCC RA: CHAPEL VS. C+MPI (TODAY)



# HPCC RA: CHAPEL VS. C+MPI

```
/* 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 & (TABSIZ-1)] ^= Ran;  
 * }  
  
MPI_Irecv(&LocalRecvBuffer, localBufferSize,  
          MPI_ANY_SOURCE, MPI_A  
while (i < 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;  
                     inmsg = LocalRecvBuffer[b  
                     LocalOffset = (inmsg & (t  
                         tparams.Glo  
                         HPCC_Table[LocalOffset] ^=  
                     } else if (status.MPI_TAG == FINISHED_TAG) {  
                        NumberReceiving--;  
                    } else  
                        MPI_Abort( MPI_COMM_WORLD,  
                                  MPI_Irecv(&LocalRecvBuffer, 1  
                                            MPI_ANY_SOURCE, MPI_A  
                } 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++;
```

Chapel Kernel

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

```
MPI_Test(&outreq, &have_done, MPI_STATUS_IGNORE);  
if (have_done) {  
    outreq = MPI_REQUEST_NULL;  
    /* HPCC_Generate_Updates localSendBuffer, localBufferSize,  
       , tparams.dtype64, (int)pe,  
       &outreq);
```

```
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_Put( tparams.dtype64, proc_count, FINISHED_TAG,  
             tparams.finish_req + proc_count);
```

Now, think about what it would take for a compiler to optimize the C+MPI code...

...or for a user to target the Cray XC's network atomics manually (and portably?)

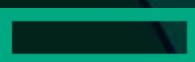
```
HPCC_Table[LocalOffset] ^= inmsg;  
}  
} else if (status.MPI_TAG == FINISHED_TAG) {  
    /* we got a done message. Thanks for playing... */  
    NumberReceiving--;  
} else {  
    MPI_Abort( MPI_COMM_WORLD, -1 );  
}  
MPI_Irecv(&LocalRecvBuffer, localBufferSize, tparams.dtype64,  
          MPI_ANY_SOURCE, MPI_ANY_TAG, MPI_COMM_WORLD, &inreq);  
}  
} while (have_done && NumberReceiving > 0);
```

```
if (status.MPI_TAG == UPDATE_TAG) {  
    MPI_Get_count(&status, tparams.dtype64, &recvUpdates);  
    bufferBase+j;  
    tparams.TableSize - 1)) -  
    GlobalStartMyProc;  
    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);
```

# OUTLINE

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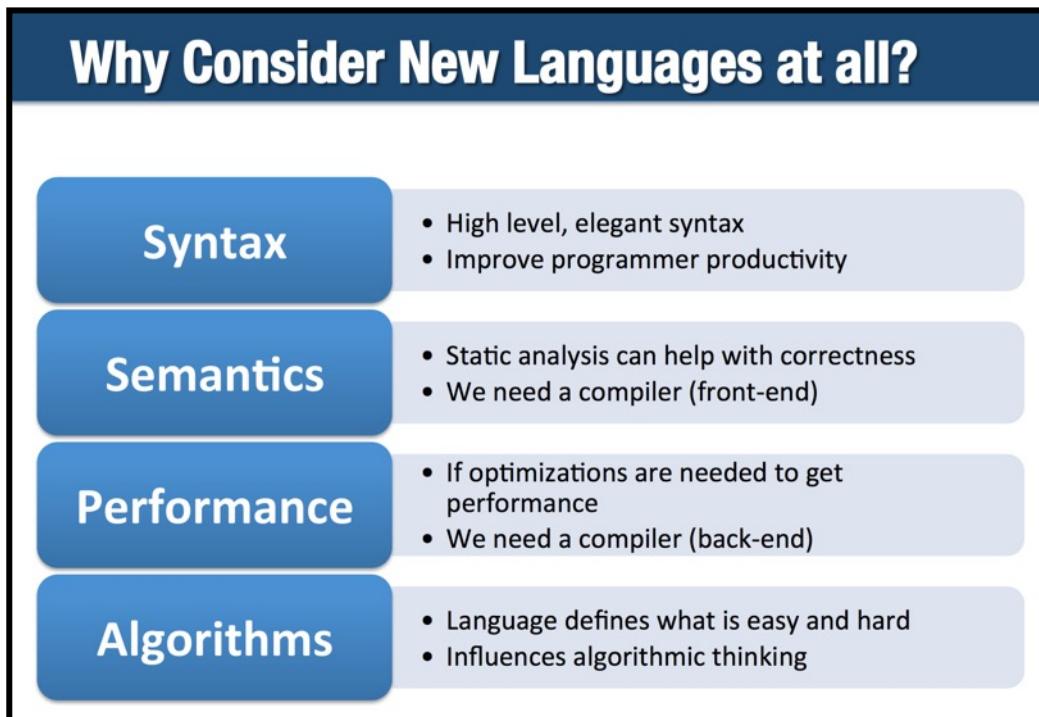
- I. Chapel by Example: HPCC RA
- II. Chapel Motivation & Context
- III. Tasking and Locality Features
- IV. Chapel Aggregators
- V. Arkouda and Aggregation
- VI. Wrap-up



# 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)  
[CHI UW 2018](#) keynote:  
[Why Languages Matter More Than Ever](#),  
used with permission]

- **And because existing languages don't fit our desires...**

# CHAPEL, RELATIVE TO OTHER LANGUAGES

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**Chapel strives to be as...**

...**programmable** as Python

...**fast** as Fortran

...**scalable** as MPI, SHMEM, or UPC

...**portable** as C

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

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



## **CHAPEL BENCHMARKS TEND TO BE CONCISE, CLEAR, AND COMPETITIVE**

## **STREAM TRIAD: C + MPI + OPENMP**

```

#include <hpcc.h>
#ifndef _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( 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) != 0) {
        if (a) HPCC_free(a);
        if (b) HPCC_free(b);
        if (c) HPCC_free(c);
        if (doIO) {
            fprintf( outFile, "Failed to allocate memory\n" );
            fclose( outFile );
        }
        return 1;
    }

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

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

```

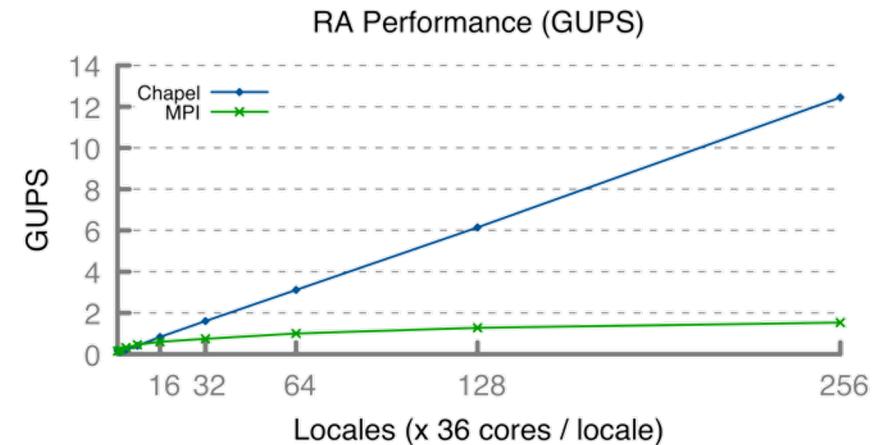
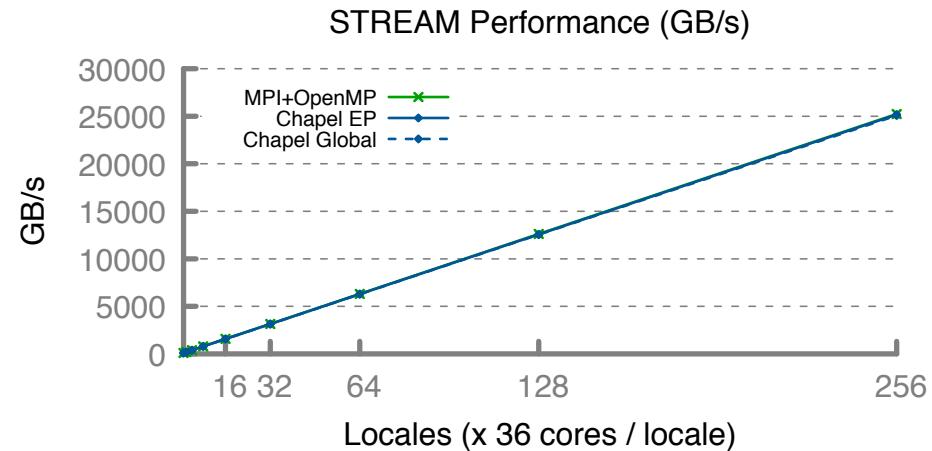
```
use BlockDist;

config const m = 1000,
          alpha = 3.0;
const Dom = {1..m} dmapped ...;
var A, B, C: [Dom] real;

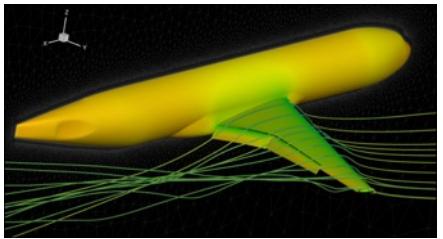
B = 2.0;
C = 1.0;

A = B + alpha * C;
```

## **HPCC RA: MPI KERNEL**

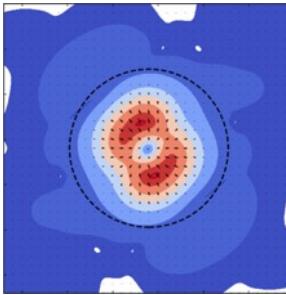


# NOTABLE CURRENT APPLICATIONS OF CHAPEL



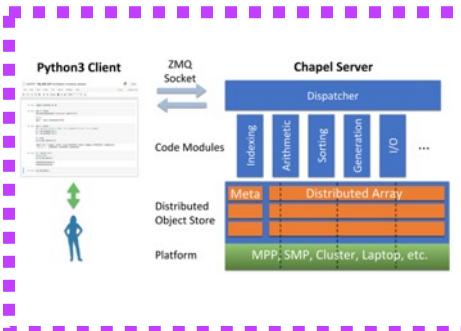
## CHAMPS: 3D Unstructured CFD

Éric Laurendeau, Simon Bourgault-Côté,  
Matthieu Parenteau, et al.  
École Polytechnique Montréal  
~120k lines of Chapel



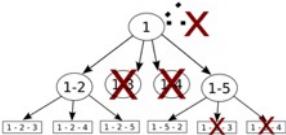
## ChplUltra: Simulating Ultralight Dark Matter

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



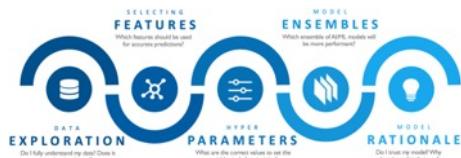
## Arkouda: NumPy at Massive Scale

Mike Merrill, Bill Reus, et al.  
US DOD  
~22k lines of Chapel



## ChOp: Chapel-based Optimization

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



## CrayAI: Distributed Machine Learning

Hewlett Packard Enterprise



## Your Project Here?

# CHAPEL'S MULTIRESOLUTION PHILOSOPHY

1. Users should be able to program at high levels of abstraction and get good performance

```
Dst = Src[Inds]; // whole-array index gather
```

2. Yet, when more control / better performance is needed, they can drop to lower levels...

```
forall (d, i) in zip(Dst, Inds) do // parallel loop-based index gather
    d = Src[i];
```

...and even lower levels, as necessary...

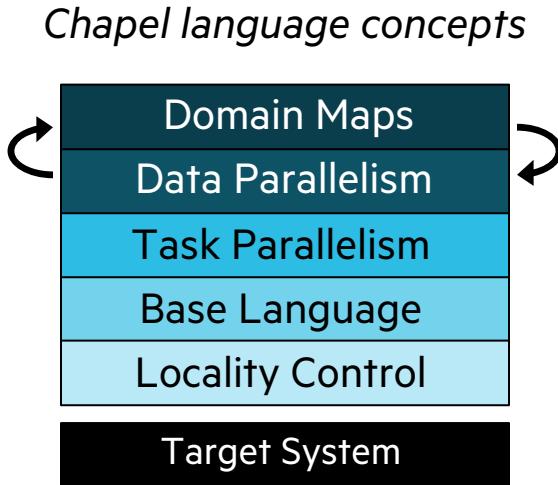
```
coforall loc in Dst.targetLocales do // explicit SPMD-style index gather
    on loc do
        forall i in Dst.localSubdomain do
            Dst.localAccess[i] = Src[Inds.localAccess[i]];
```

...where “calling out to C / CUDA / MPI / etc.” is effectively the lowest level

3. Chapel builds its higher-level abstractions in terms of the lower-level ones to guarantee compatibility

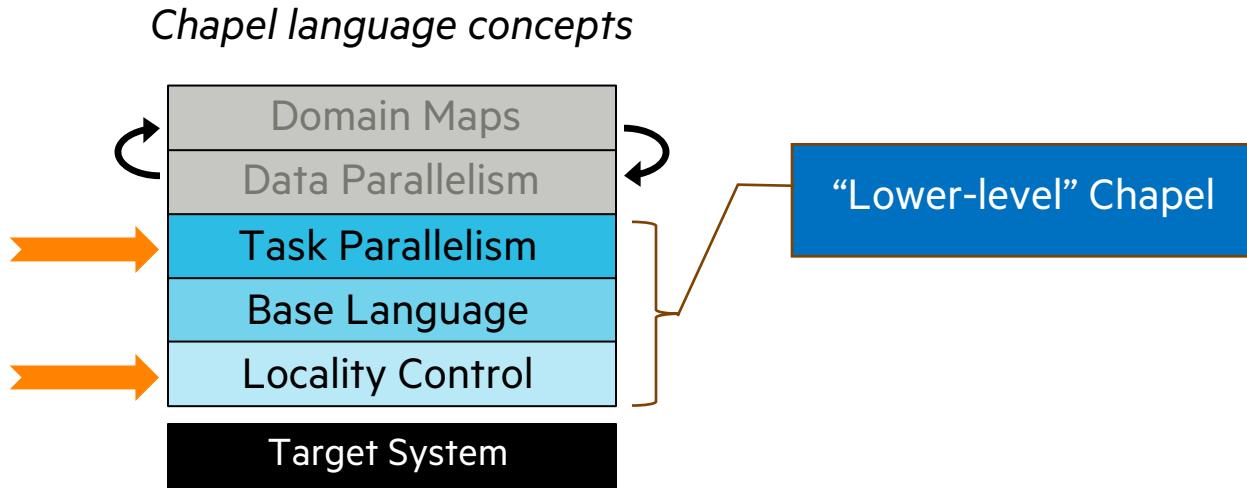
# CHAPEL'S MULTIRESOLUTION FEATURE STACK

---



# CHAPEL’S “LOWER-LEVEL” FEATURES

---



## **TASKING AND LOCALITY FEATURES**

# CHAPEL, ASYNCHRONY, AND THIS TALK

---

- Chapel tasks are asynchronous in sense that they are:
  - launched dynamically
  - managed by the runtime
  - no pre-determined start or end conditions
- Asynchrony gives expressiveness benefits in that any parallel pattern can be expressed
- In practice, many patterns use groups of tasks doing similar things
  - such as the coforall examples in this talk
- Asynchrony improves performance by spreading network load and avoids synchronization bottlenecks
  - aggregators in this talk benefit highly from this



# CHAPEL TERMINOLOGY: LOCALES

- Locales can run tasks and store variables
  - Think “compute node” on a parallel system
  - User specifies number of locales on executable’s command-line

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

**Locales** array:



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

# TASK-PARALLEL “HELLO WORLD”

helloTaskPar.chpl

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

‘here’ refers to the locale on which we’re currently running

how many processing units (think “cores”) does my locale have?

what’s my locale’s name?

# TASK-PARALLEL “HELLO WORLD”

helloTaskPar.chpl

```
const numTasks = here.numPUs();
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.numPUs();
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.numPUs();
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.numPUs();
        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.numPUs();  
        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  
...
```

# DIFFERENCES BETWEEN CHAPEL AND TRADITIONAL PGAS / SHMEM

---

## 1. Chapel supports a post-SPMD execution model

- **traditional PGAS:** all PEs/ranks/threads start by executing ‘main’
- **Chapel:** a single task executes ‘main’ on locale 0 and additional parallelism\* is introduced from there

(\* = local or distributed)

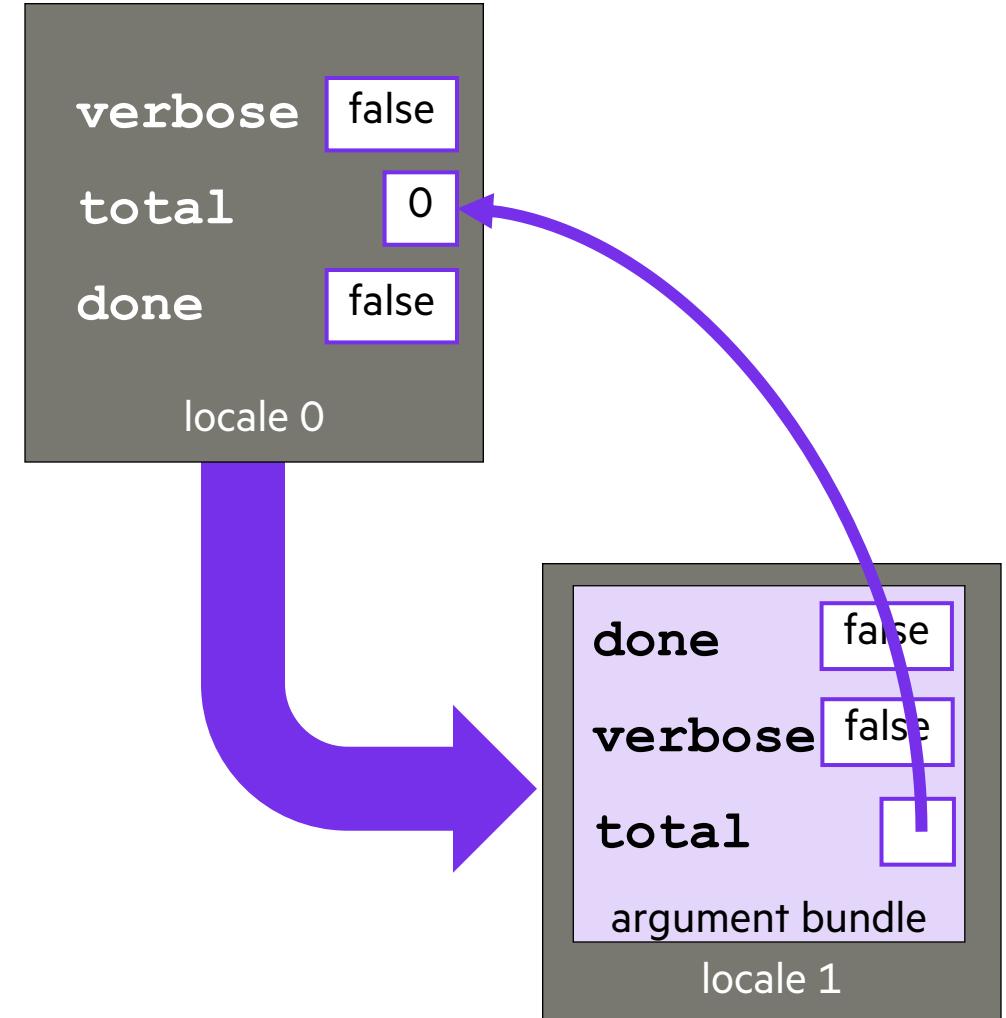


# CHAPEL'S PARTITIONED GLOBAL NAMESPACE

onClause.chpl

```
const verbose = false;
var total = 0,
    done = false;

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



# DIFFERENCES BETWEEN CHAPEL AND TRADITIONAL PGAS / SHMEM

---

1. Chapel supports a post-SPMD execution model
  - **traditional PGAS:** all ranks/threads/PEs start by executing ‘main’
  - **Chapel:** a single task executes ‘main’ on locale 0 and additional parallelism\* is introduced from there  
(\* = local or distributed)
  
2. Chapel’s partitioned global address space is also post-SPMD
  - **traditional PGAS:** “I have a variable named ‘x’, so you must too, and therefore I can refer to yours”
  - **Chapel:** “I see variable ‘x’ in my lexical scope, so I can refer to it, whether it’s local or remote”

*One outcome of these differences is that Chapel feels much more like traditional programming*



# BULK COMMUNICATION IN CHAPEL: A TOOL FOR MANUAL AGGREGATION

bulkComm.chpl

```
var Buff: [0..<buffSize] real;  
on Locales[1] {  
    var LocBuff = Buff;  
    processData(LocBuff);  
    Buff = LocBuff;  
}
```

allocate an array on locale 0

move computation to locale 1

bulk 'get' from remote array

bulk 'put' to remote array

# **CHAPEL AGGREGATORS**

# BALE INTRODUCTION

---

- Bale is a collection of mini applications and aggregation libraries
  - Chapel has several ports of Bale applications, including index gather
  - We use Bale to evaluate the productivity of our aggregators and to compare performance to SHMEM
- From the description in <https://github.com/jdevinney/bale>, Bale is a:  
*“Vehicle for discussion for parallel programming productivity. The bale effort attempts to:*
  - demonstrate some challenges of implementing interesting (i.e. irregular) scalable distributed parallel applications.*
  - demonstrate an approach (aggregation) to achieve high performance for the internode communication in such applications*
  - explore concepts that make it easier to write, maintain, and get top performance from such applications**We use bale to evolve our thinking on parallel programming in the effort to make parallel programming easier, more productive, and more fun. Yes, we think making it fun is a worthy goal!”*



# BALE INDEX GATHER IN CHAPEL

```
use BlockDist, Random;

const numTasks = numLocales * here.maxTaskPar;
config const N = 1000000, //number of updates per task
      M = 10000;      //number of entries in the table per task

const D = newBlockDom(0..<M*numTasks);
var Src: [D] int = D;
const UpdatesDom = newBlockDom(0..<N*numTasks);
var Dst, Inds: [UpdatesDom] int;

fillRandom(Inds, min=0, max=Src.size);

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

# BALE INDEX GATHER KERNEL IN CHAPEL: NAÏVE VERSION

---

```
// Naive index gather: Dst = Src[Inds];  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

‘Src’ is a distributed array with  
*numEntries* elements

‘Dst’ and ‘Inds’ are distributed arrays with  
*numUpdates* elements

# BALE INDEX GATHER KERNEL IN CHAPEL: NAÏVE VERSION

```
// Naive index gather: Dst = Src[Inds];  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

Gets lowered roughly to...

```
coforall loc in Dst.targetLocales do ← A concurrent loop over the compute nodes  
    on loc do  
        coforall tid in 0..<here.maxTaskPar do ← A nested concurrent loop over each node's cores  
            for idx in myInds(loc, tid, ...) do ← A serial loop to compute each task's chunk of gathers  
                D[idx] = Src[Inds[idx]];
```

# BALE INDEX GATHER KERNEL IN CHAPEL: NAÏVE VERSION

```
// Naive index gather: Dst = Src[Inds];  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

Gets lowered roughly to...

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

But, for a parallel loop with no data dependencies,  
why perform these high-latency operations serially?

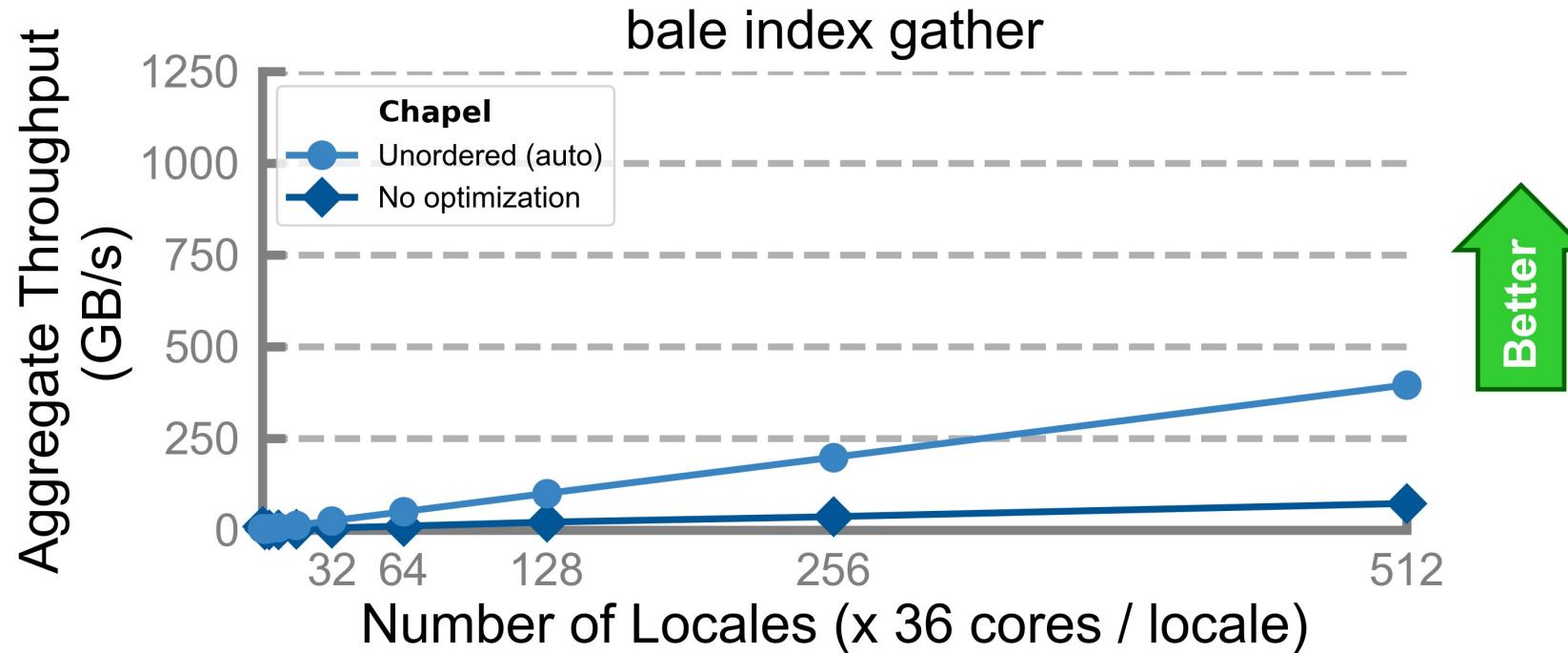
```
for idx in myInds(loc, tid, ...) do  
    unorderedCopy(D[idx], Src[Inds[idx]]);  
    unorderedCopyTaskFence();
```

So, our compiler rewrites the inner loop  
to perform the ops asynchronously

- Implemented by Michael Ferguson, 2019

# BALE INDEX GATHER KERNEL IN CHAPEL: NAÏVE VERSION

```
// Naive index gather: Dst = Src[Inds];  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```



# BALE INDEX GATHER KERNEL IN CHAPEL: AGGREGATOR VERSION

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

```
use CopyAggregation; → 'use' the module providing the aggregators

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

To use it, we simply replace  
the assignment with 'agg.copy'

'use' the module providing the aggregators

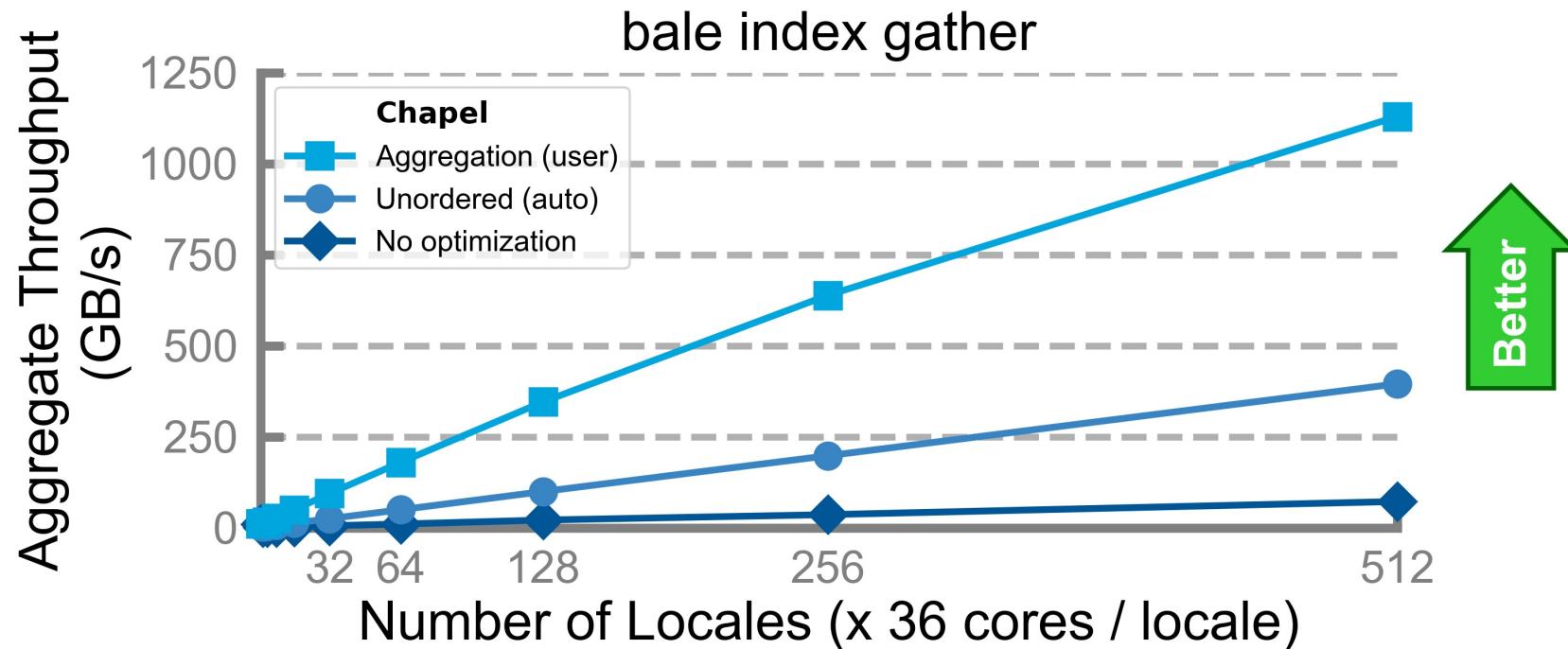
Give each task a “source aggregator”, *agg*, which  
aggregates remote ‘gets’ locally, then performs them

As the aggregator’s buffers fill up, it communicates the operations  
to the remote locale, automatically and asynchronously

# BALE INDEX GATHER KERNEL IN CHAPEL: AGGREGATOR VERSION

```
use CopyAggregation;

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



# CAN WE AUTOMATE AGGREGATION?

**Q:** Is there an opportunity for the compiler to introduce aggregators automatically?

```
// Naive index gather: Dst = Src[Inds];  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

user writes straightforward code  
compiler optimizes as:

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

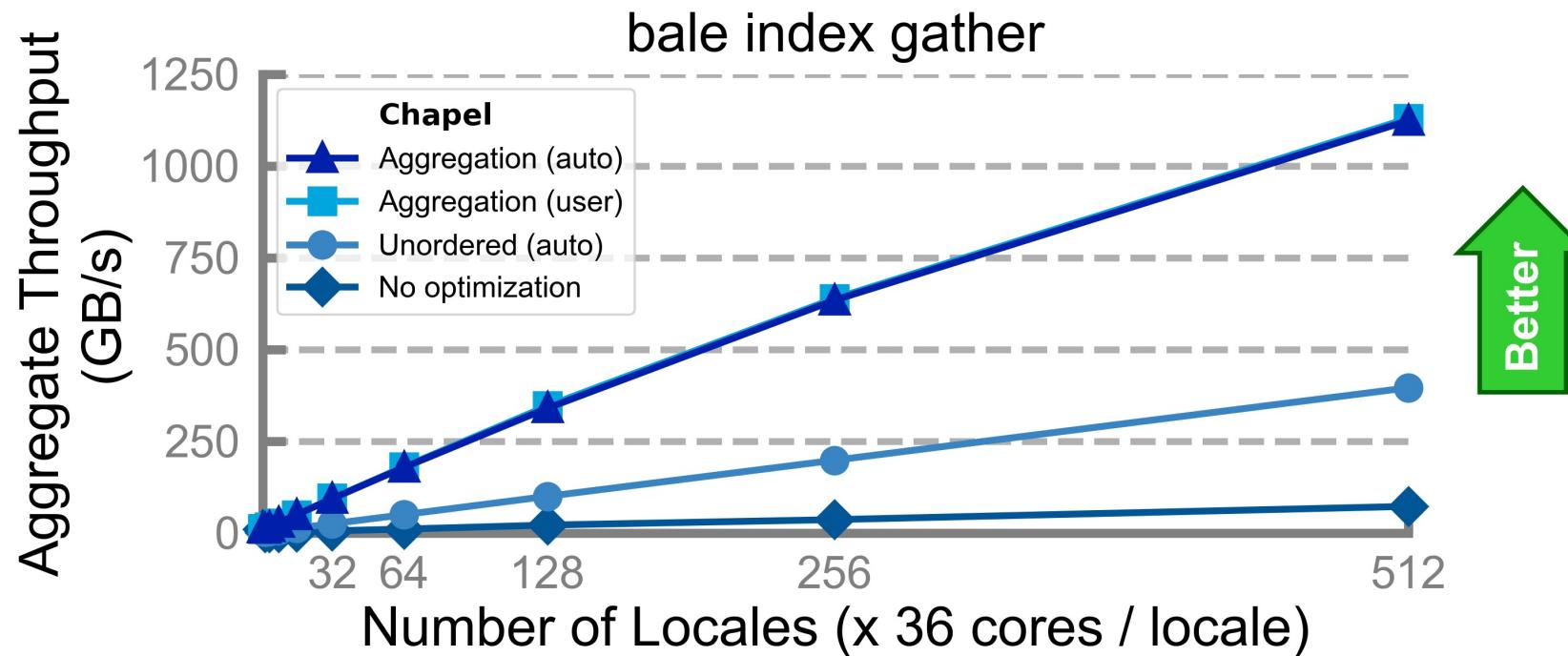
**A:** In many cases, yes

- developed by Engin Kayraklıoglu, 2021
- combines previous ‘unordered’ analysis with a new locality analysis of RHS/LHS expressions
- for details, see Engin’s LCPC 2021 paper: <https://lcpc2021.github.io/>

# AUTO-AGGREGATION: IMPACT

- As a result, the naïve version can now compete with the user-written aggregators

```
// Naive index gather: Dst = Src[Inds];  
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```



# BALE INDEX GATHER: CHAPEL VS. EXSTACK VS. CONVEYORS

## Elegant SHMEM version

```
for (i = 0; i < N; i++)  
    shmem_get(&target[i], &table[index[i]], sizeof(long), index[i] % NPES);
```

## Exstack version

```
while( exstack_proceed(ex, (i==l_num_req)) ) {  
    i0 = i;  
    while(i < l_num_req) {  
        l_idx = pckindx[i] >> 16;  
        pe = pckindx[i] & 0xfffff;  
        if(!exstack_push(ex, &l_idx, 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=10; j<i; j++) {  
            fromth = pckindx[j] & 0xfffff;  
            exstack_pop_thread(ex, &idx, (uint64_t)fromth);  
            tgt[j] = idx;  
        }  
        lgp_barrier();  
    }  
}
```

## Conveyors version

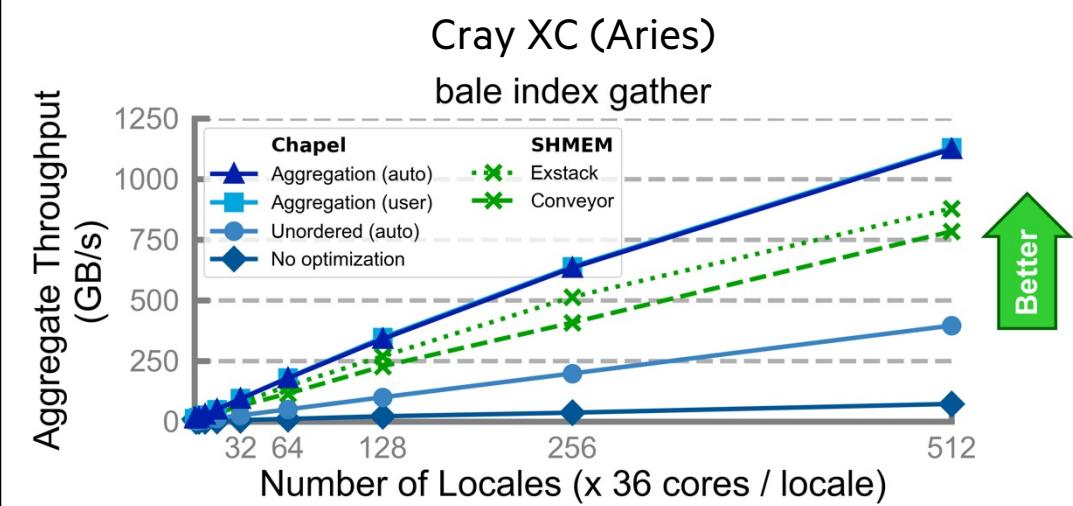
```
i = 0;  
while (more = convey_advance(requests, (i == l_num_req)),  
      more | convey_advance(replies, !more)) {  
  
    for ( ; i < l_num_req; i++) {  
        pkg.idx = i;  
        pkg.val = pckindx[i] >> 16;  
        pe = pckindx[i] & 0xfffff;  
        if (!convey_push(requests, &pkg, pe)) 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')

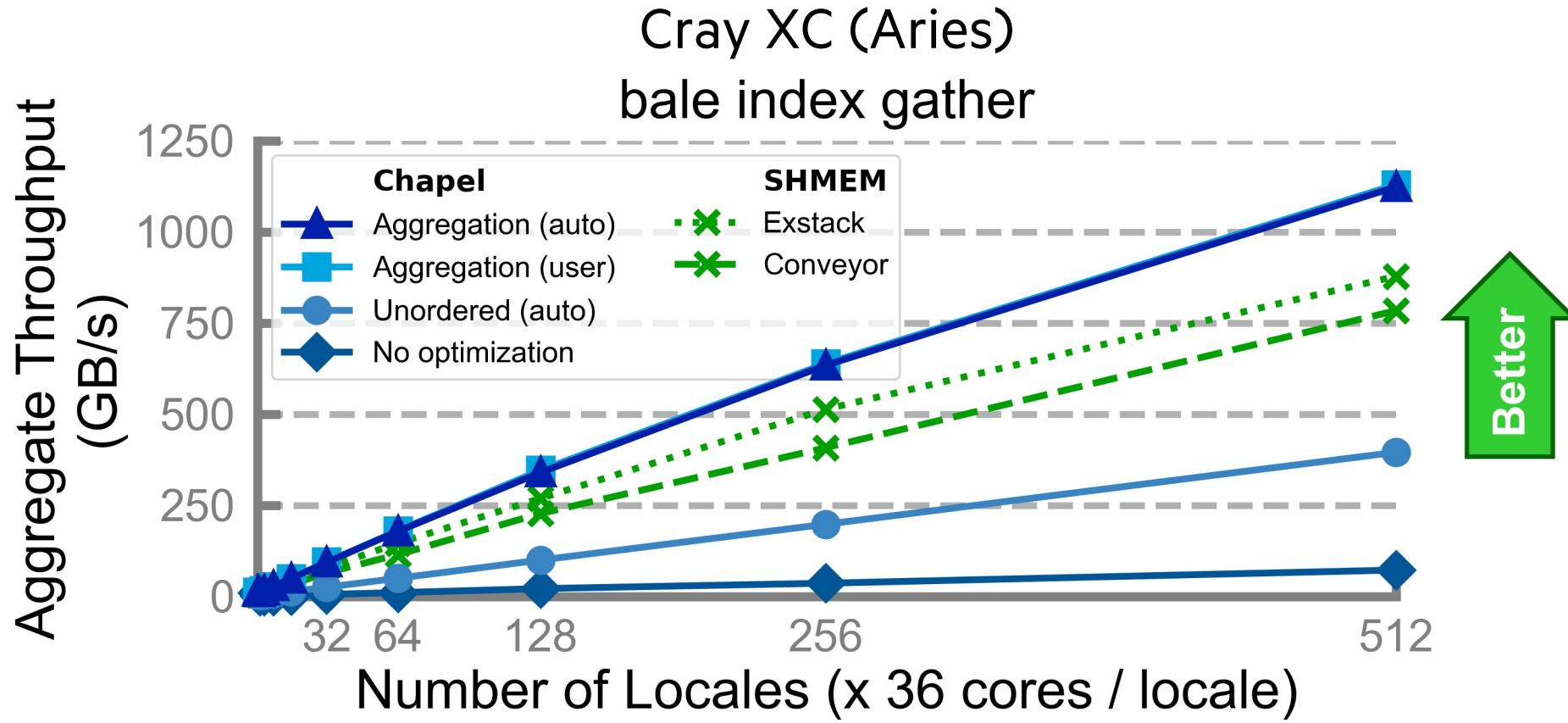
```
forall (d, i) in zip(Dst, Inds) do  
    d = Src[i];
```

## 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]);
```



# BALE INDEX GATHER: CHAPEL VS. EXSTACK VS. CONVEYORS



# IMPLEMENTING CHAPEL'S AGGREGATORS

---

- Chapel's aggregators are implemented as Chapel source code
  - no language or compiler changes were required
  - initial implementation only relied on high-level features
    - current optimized version calls into lower-level put/get routines
  - Relies upon:
    - standard language features:
      - OOP: records, initializers, de-initializers
      - arrays
      - access to C-level pointers and dereferences
    - Chapel features that you've seen:
      - global namespace
      - task-local variables
  - ~100 lines of reasonably straightforward code to implement SrcAggregator
    - (~420 lines for entire 'CopyAggregation' module)



# INITIAL SRC AGGREGATOR IMPLEMENTATION: EXCERPTS

```
record SrcAggregator {  
    type elemType;  
    var dstAddrs, srcAddrs: [LocaleSpace] [0..<bufferSize] addr;  
    var bufferIdxs: [LocaleSpace] int;  
    ...  
    proc flushBuffer(loc: int, ref bufferIdx) {  
        var srcVals: [0..<bufferIdx] elemType;  
  
        on Locales[loc] {  
            const locSrcAddrs = srcAddrs[loc] [0..<bufferIdx];  
            var locSrcVals: [0..<bufferIdx] elemType;  
            ... // fill the locSrcVals array  
            srcVals = locSrcVals;  
        }  
        ... // assign the srcVals to the dstAddrs  
    }  
}
```

Arrays for buffering per-locale src/dest addresses

time to flush a buffer?

allocate a landing spot for the remote src values

move to the remote node to buffer src vals and copy them back

Bulk array copy to 'get' the src addresses

Declare a buffer to store the local src vals and fill it

Bulk array copy to 'put' src values back to original locale

Store src vals to dest addresses

# INITIAL SRC AGGREGATOR IMPLEMENTATION: EXCERPTS

```
record SrcAggregator {
    type elemType;
    var dstAddrs, srcAddrs: [LocaleSpace] [0..<bufferSize] addr;
    var bufferIdxs: [LocaleSpace] int;
    ...
    proc flushBuffer(loc: int, ref bufferIdx) {
        var srcVals: [0..<bufferIdx] elemType;

        on Locales[loc] {
            const locSrcAddrs = srcAddrs[loc][0..<bufferIdx];
            var locSrcVals: [0..<bufferIdx] elemType;
            ... // fill the locSrcVals array
            srcVals = locSrcVals;
        }
        ... // assign the srcVals to the dstAddrs
    }
}
```

Bulk array copy to 'get'  
the src addresses

Bulk array copy to 'put'  
src values back to original locale

# CHAPEL AGGREGATORS: ATTRACTIVE PROPERTIES

---

- More flexible than traditional aggregators:
  - **traditional aggregators:** like barriers or collectives, tend to assume everyone is involved and quasi-lockstep
  - **Chapel aggregators:** Chapel's post-SPMD nature relaxes traditional BSP constraints
    - tasks communicate with remote locales asynchronously, once a given buffer fills up
    - any subset of tasks/locales can utilize aggregators that target any locales *without those locales being involved*
- User-level tasks make the implementation efficient
  - Chapel leverages Sandia's Qthreads
- Performance is competitive with conventional techniques



# SOUNDS GREAT, WHAT'S THE CATCH?

**Q:** Clean code, competitive performance and scalability, no modifications to the language or compiler...  
...so, what's the catch?

**A:** Not a 'catch' per se, but currently, Chapel's aggregators only support copy-style and atomic operations

- Ultimately, want/need to support general operations ("user-defined aggregators")
  - In principle, not so different from the existing ones
  - **Limiting factor:** These would most naturally be expressed with first-class functions (FCFs)
    - ...but Chapel's support for FCFs is currently a bit weak
- That said, many interesting computations can be written with copy-style aggregation...
  - ...like Arkouda!



# **ARKouda AND AGGREGATION**

# MOTIVATION FOR ARKOUDA

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

- ...HPC-scale data science problems to solve
- ...a bunch of Python programmers
- ...access to HPC systems

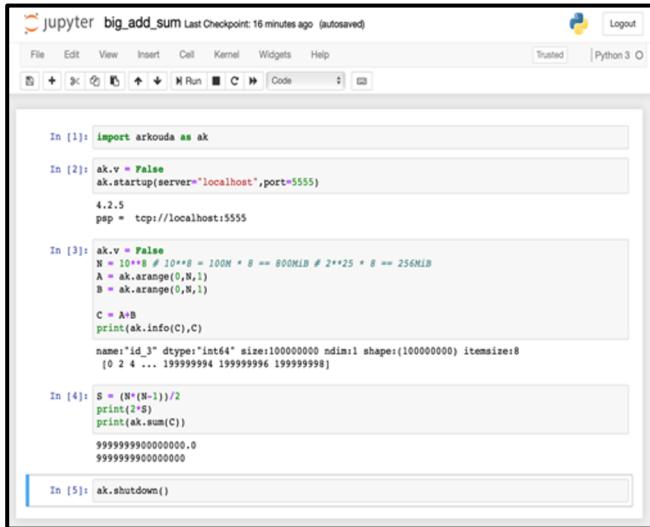


How will you leverage your Python programmers to get your work done?



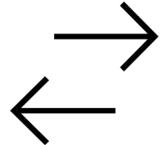
# ARKOUDA'S HIGH-LEVEL APPROACH

## Arkouda Client (written in Python)



A screenshot of a Jupyter Notebook interface. The code cell In [1] imports the Arkouda library. In [2] starts the Arkouda server on localhost port 5555. In [3] creates two large arrays A and B of size 100M x 8. In [4] performs a dot product C = A\*B and prints its memory information. In [5] prints the sum of array S.

```
In [1]: import arkouda as ak
In [2]: ak.v = False
ak.startup(server="localhost",port=5555)
4.2.5
psp = tcp://localhost:5555
In [3]: ak.v = True
N = 10**8 * 10**8 = 100M * 8 == 800MB # 2**25 * 8 == 256MB
A = ak.arange(0,N,1)
B = ak.arange(0,N,1)
C = A*B
print(ak.info(C),C)
name:id_3 dtype:int64 size:1000000000 ndim:1 shape:(100000000) itemsize:8
[0 2 4 ... 199999994 199999996 199999998]
In [4]: S = (N*(N-1))/2
print(2*S)
print(ak.sum(C))
9999999900000000.0
9999999900000000
In [5]: ak.shutdown()
```



## Arkouda Server (written in Chapel)



User writes Python code in Jupyter,  
making familiar NumPy/Pandas calls



# ARKOUDA SUMMARY

## What is it?

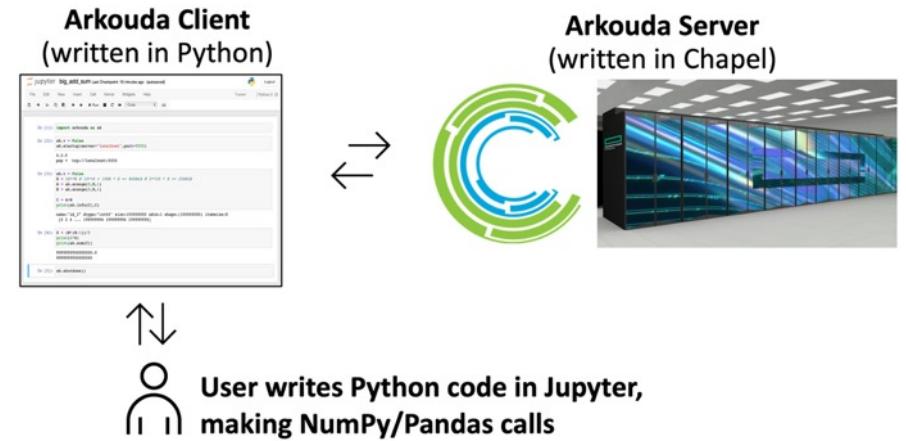
- A Python library supporting a key subset of NumPy and Pandas for Data Science
  - Uses a Python-client/Chapel-server model to get scalability and performance
  - Computes massive-scale results (multi-TB-scale arrays) within the human thought loop (seconds to a few minutes)
- ~22k lines of Chapel, largely 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>

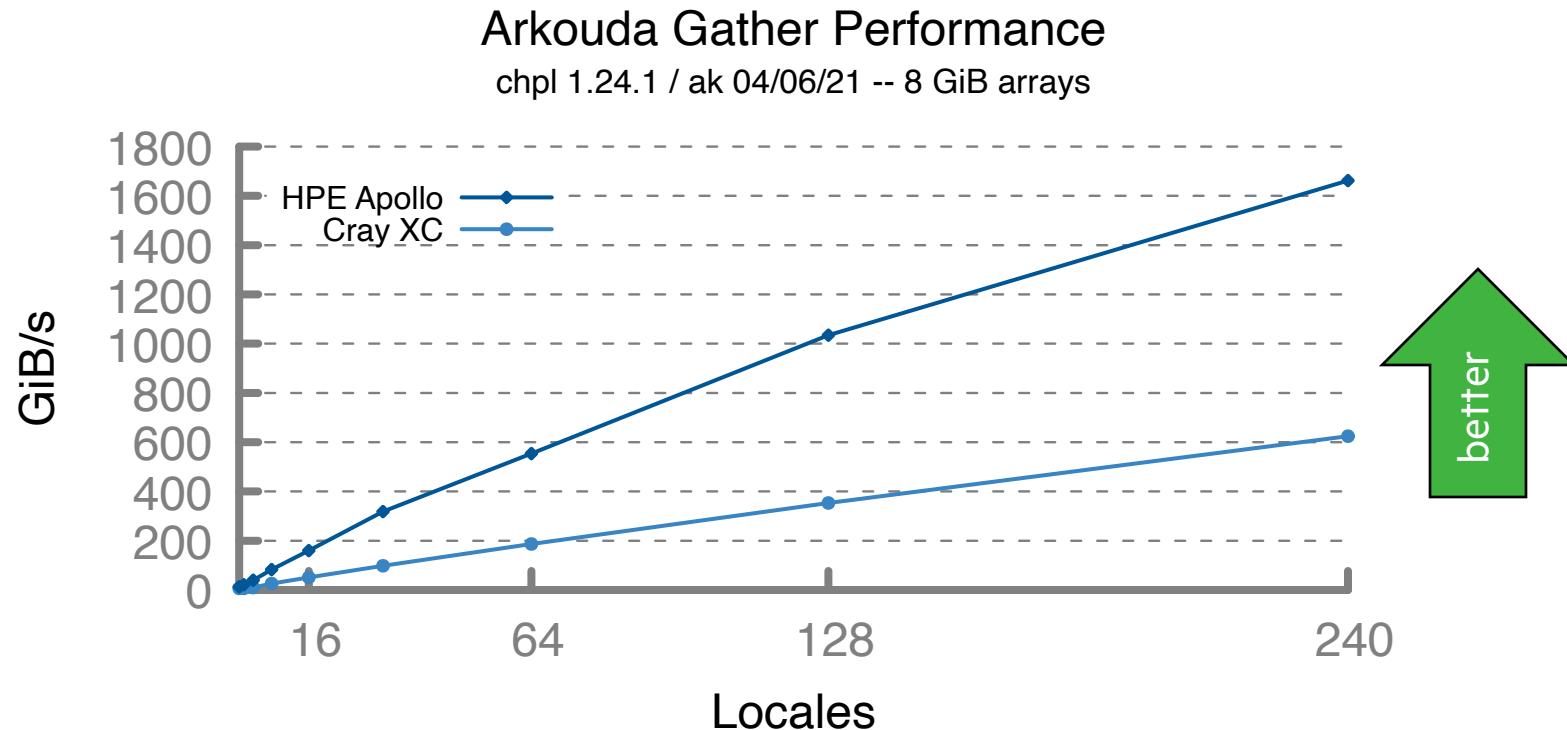
## Why Chapel?

- high-level language with performance and scalability
- close to Pythonic
  - enabled writing Arkouda rapidly
  - doesn't repel Python users who look under the hood
- ports from laptop to supercomputer



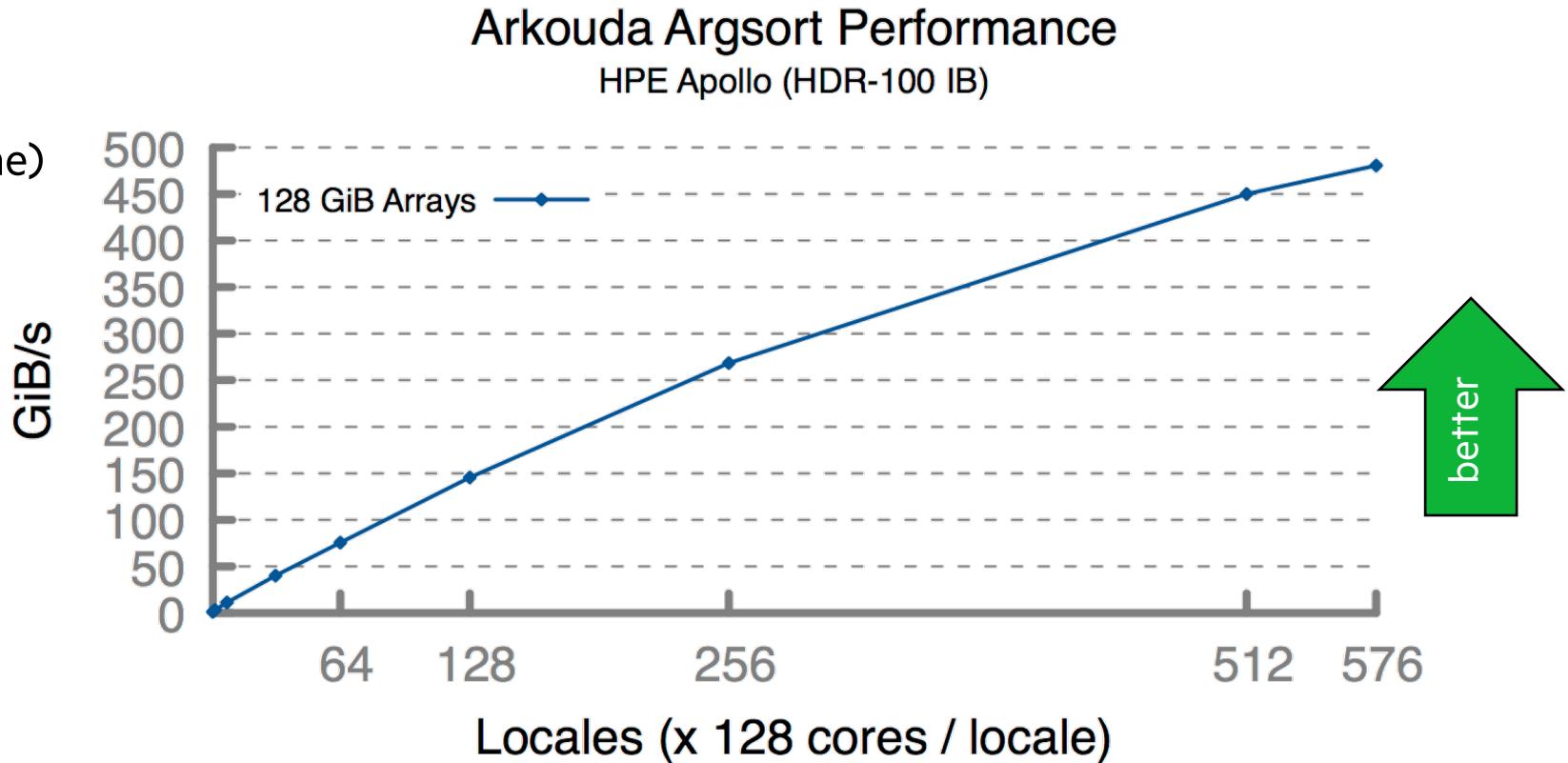
## ARKOUDA GATHER

- For Arkouda's gather kernel, Chapel performance on a recent HPE Apollo system is well ahead of XC
  - These timings were taken in April 2021
  - System-level bugs hurt reference SHMEM performance, so no direct comparisons here



# ARKOUDA ARG SORT AT SCALE

- Ran on a large Apollo system, summer 2021
  - 73,728 cores of AMD Rome
  - 72 TiB of 8-byte values
  - 480 GiB/s (2.5 minutes elapsed time)
  - ~100 lines of Chapel code



**Close to world-record performance—quite likely a record for performance/SLOC**



The background features a series of wavy, overlapping diagonal stripes in a teal color. The stripes are oriented from the top-left towards the bottom-right. They are layered, creating a sense of depth. The lighting is soft, highlighting the curves of the waves and the edges of the stripes.

**WRAP-UP**

# CHAPEL RESOURCES

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

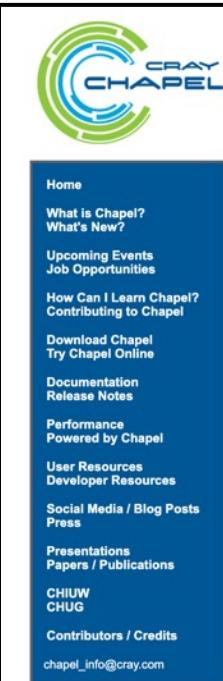
- (points to all other resources)

## Social Media:

- Twitter: [@ChapelLanguage](#)
- Facebook: [@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

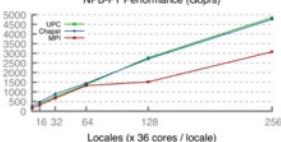
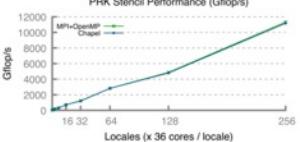
**Chapel Characteristics**

- 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:



- 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

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## 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\*](#) (CHIUW 2021), June 2021
- [\*Chapel Release Notes\*](#) — current version 1.24, April 2021

## Arkouda:

- Bill Reus's CHIUW 2020 keynote talk: <https://chapel-lang.org/CHIUW2020.html#keynote>
- Arkouda GitHub repo and pointers to other resources: <https://github.com/Bears-R-Us/arkouda>

## CHAMPS:

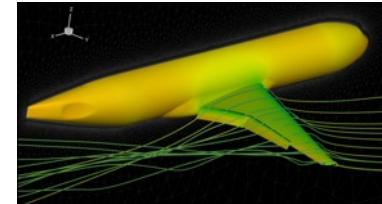
- Eric Laurendeau's CHIUW 2021 keynote talk: <https://chapel-lang.org/CHIUW2021.html#keynote>
  - two of his students also gave presentations at CHIUW 2021, also available from the URL above
- Another paper/presentation by his students at <https://chapel-lang.org/papers.html> (search “Laurendeau”)



# SUMMARY

## Chapel is designed for productive parallel programming at scale

- recent users have reaped these benefits in large applications



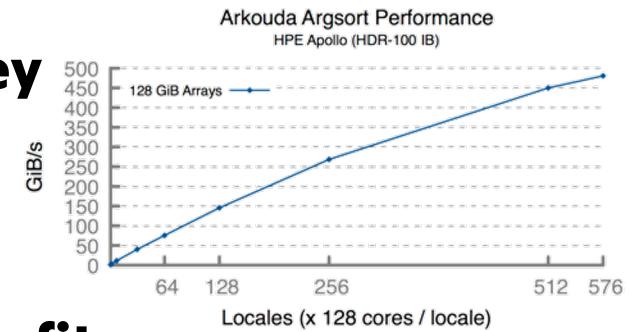
## Though PGAS in nature, Chapel avoids SPMD / BSP assumptions

- parallelism is expressed in the source code starting from a single task
- lexical scoping simplifies PGAS-based communication
- the net result is a far more approachable distributed parallel language

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

## For gather/scatter/sort in Arkouda and Bale, copy aggregators are key

- Chapel's are implemented concisely and elegantly within the language
- performance rivals that of Exstack / Conveyors



## Chapel's design and language-based nature provide optimization benefits

- e.g., automatic asynchronous operations and automatic aggregation (as in Arkouda / Bale)



# THANK YOU

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

