

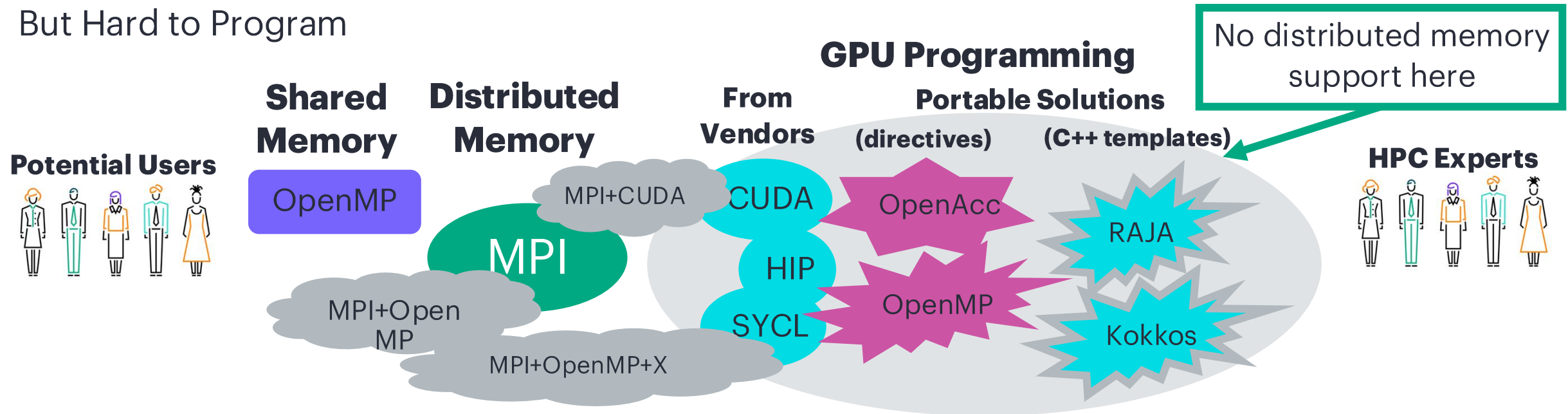
# Productive Parallel Programming with Chapel and Arkouda

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FOSDEM 2026  
February 1, 2026

# Parallel Systems are Easy to Find...

But Hard to Program



**All are effective, powerful, essential, and tested technologies!**

- ...but
  - all of these are based on C/C++/Fortran
  - some paradigms haven't changed in decades (i.e. MPI)
  - mixing parallel hardware requires multiple frameworks
  - higher-level abstractions often exchange performance and control for ease-of-use

As a result, HPC has a high barrier to entry

# An Alternative for Productive Parallel Programming

**Chapel:** A modern parallel programming language

- Portable & scalable
- Open-source & collaborative
  - An HPSF / Linux Foundation project



## Goals:

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



# Productive Parallel Programming

- Imagine a language that is as...
  - ...**readable and writeable** as Python
  - ...**fast** as Fortran / C / C++ / Rust
  - ...**scalable** as MPI / SHMEM
  - ...**GPU-ready** as CUDA / HIP / OpenMP / Kokkos / OpenCL / OpenACC / ...
  - ...**portable** as C
  - ...**fun** as [your favorite language]

**This is the motivation for Chapel**

# Stream Triad: C + MPI + OpenMP vs Chapel

```
#include <hpc.h>
use BlockDist;

config const n = 1_000_000,
              alpha = 0.01;

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

B = 2.0;
C = 1.0;

A = B + alpha * C;
```

```
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 ) {
    free(c);
    free(b);
    free(a);
```

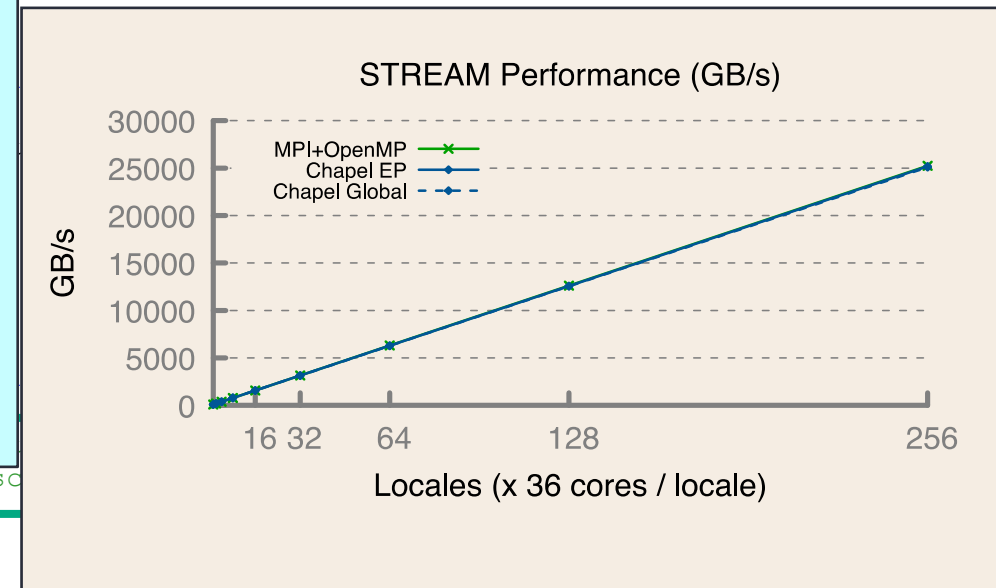
```
File, "Failed to allocate memory (%d).\n", VectorSize );
file );
```

```
a[j] = b[j] + sc
```

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

```
return 0;
```

```
}
```



# HPCC RA: C + MPI vs Chapel

/\* 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;
```

/\* 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_STATUS_IGNORE);
```

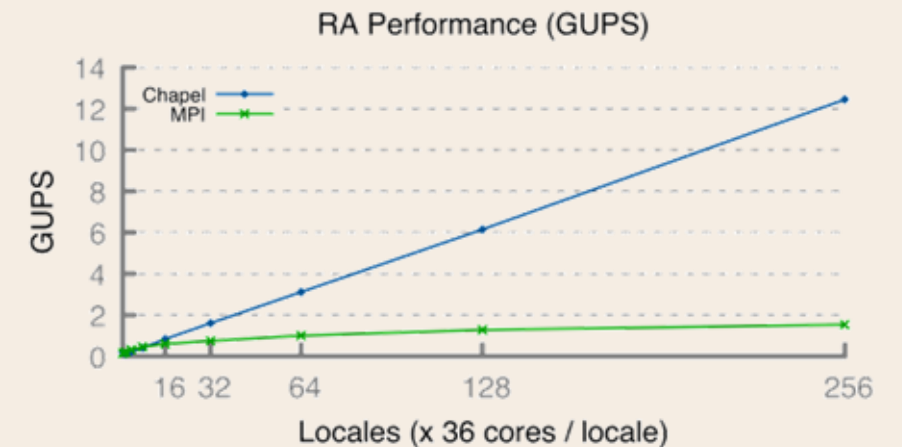
```
LocalSendBuffer, localBufferSize,
es);
beUpdates, tparams.dtype64, (int)pe,
M_WORLD, &outreq);
```

```
< tparams.NumProcs ; ++proc_count) {
pc) { 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,
```

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

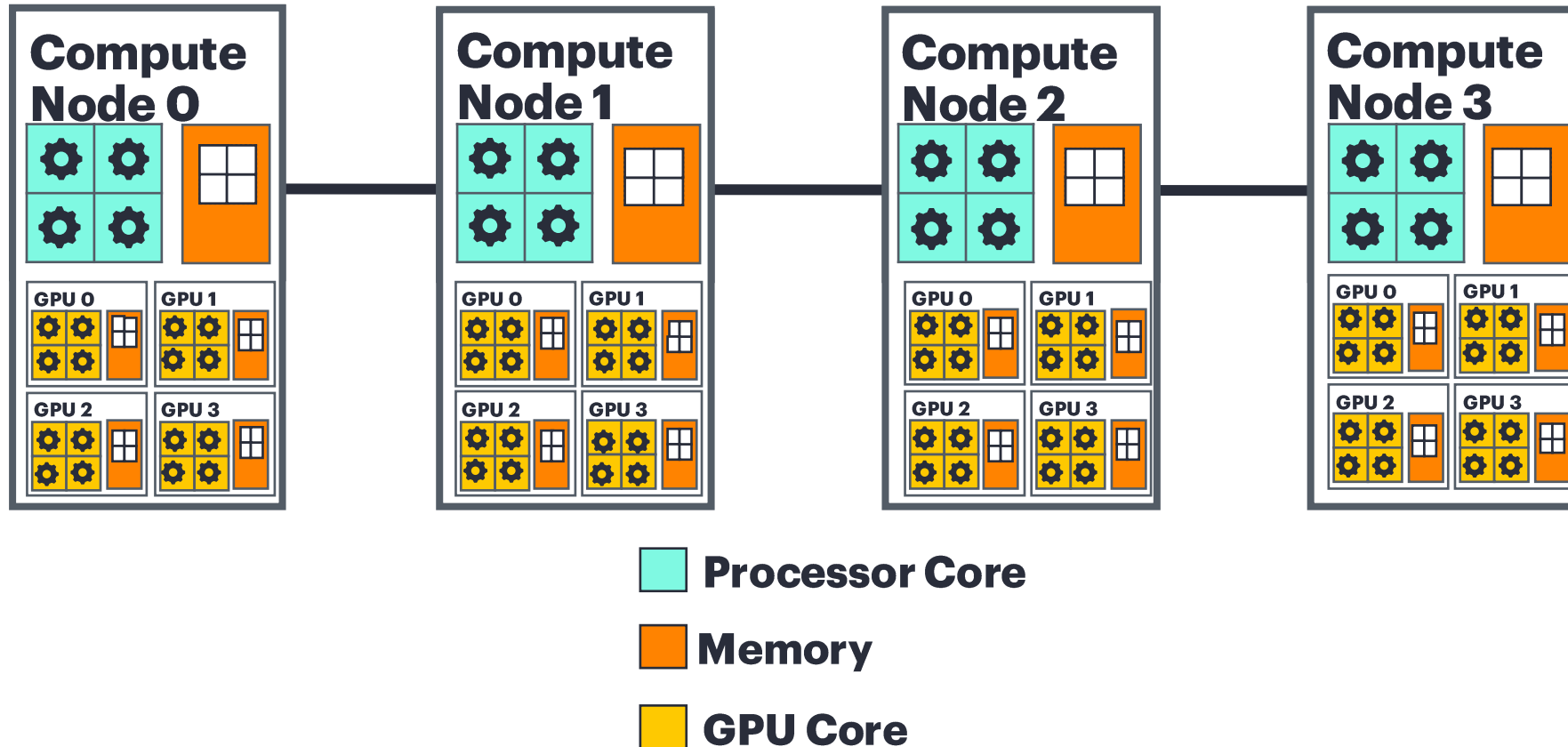
```
if (pendingupdates < maxpendingupdates) {
  Ran = (Ran << 1) ^ (((s64Int) Ran < ZERO64B ? POLY : ZERO64B);
  GlobalOffset = Ran & (tparams.TableSize-1);
  if ( GlobalOffset < tparams.Top)
    WhichPe = ( GlobalOffset / (tparams.MinLocalTableSize + 1) );
  else
    WhichPe = ( (GlobalOffset - tparams.Remainder) /
tparams.MinLocalTableSize );
  if (WhichPe == tparams.MyProc) {
    LocalOffset = (Ran & (tparams.TableSize - 1)) -
tparams.GlobalStartMyProc;
    HPCC_Table[LocalOffset] ^= Ran;
  } else 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.dty
MPI_ANY_SOURCE, MPI_ANY_TAG, MPI_COMM_WORLD, &
}
} while (have_done && NumberReceiving > 0);
```



# Key Language Features

# Key Concerns for Scalable Productive Parallel Computing

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



# Basic Features For Locality

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

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

```
for loc in Locales {
```

```
  on loc {
```

```
    var B = A;
```

```
    for gpu in loc.gpus {
```

```
      on gpu {
```

```
        var C = B;
```

```
      }
```

```
    }
```

```
  }
```

```
}
```

All Chapel programs start with one task on locale 0

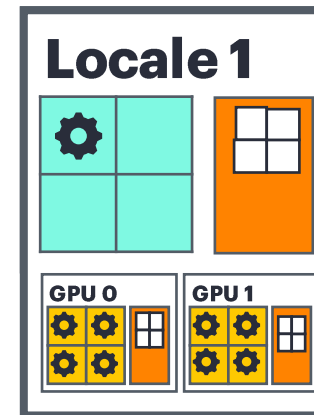
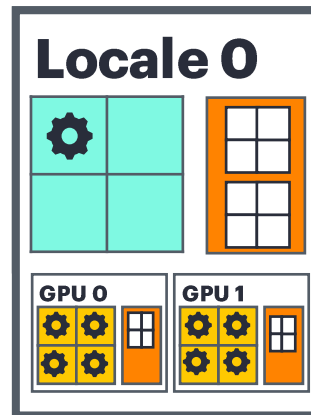
Variables are stored using the current task's local memory

A serial loop for each of the program's locales

on-clauses move the computation to the target locale

Remote variables can be accessed directly

This is distributed *serial* computation



# Locality + Parallelism

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

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

```
coforall loc in Locales {
```

```
  on loc do cobegin {
```

```
    var B = A;
```

```
    coforall gpu in loc.gpus {
```

```
      on gpu {
```

```
        var C = AB;
```

```
      }
```

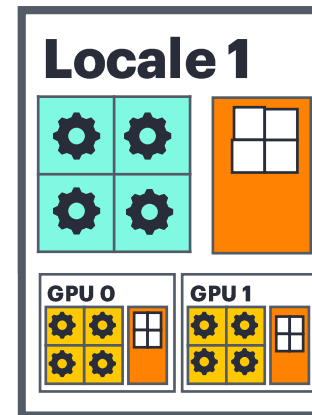
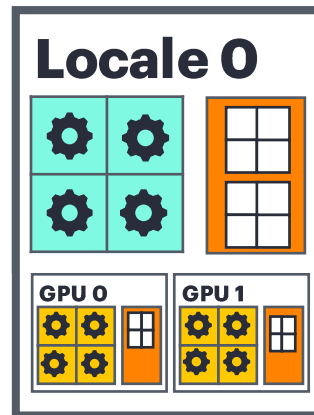
```
    }
```

```
  }
```

```
}
```

The coforall loop creates a parallel task per **iteration**

The cobegin statement creates a parallel task per **child statement**



This is distributed *parallel* computation

# One Loop To Rule Them All

```
proc increment(Arr) {  
  var Res: Arr.type;  
  forall i in Arr.domain {  
    Res[i] = Arr[i] + 1;  
  }  
  return Res;  
}
```

equivalently

```
proc increment(Arr) {  
  return Arr + 1;  
}
```

The forall loop will invoke the parallel iterator for its iterand expression, in this case the indices of Arr

```
var myLocalArr: [1..10] int;  
increment(myLocalArr);
```

Single node parallel computation

```
use BlockDist;  
var myDistributedArr = blockDist.createArray({1..10}, int);  
increment(myDistributedArr);
```

Distributed parallel computation

```
on here.gpu[0] {  
  var myGpuArr: [1..10] int;  
  increment(myGpuArr);  
}
```

Single GPU parallel computation

# One Loop To Rule Them All + Interoperability

```
extern {  
    #include <stdint.h>  
    static int64_t increment_c(int64_t X) { return X + 1; }  
}
```

It doesn't even have to be "external"!

```
proc increment(Arr) {  
    var Res: Arr.type;  
    forall i in Arr.domain {  
        Res[i] = increment_c(Arr[i]);  
    }  
    return Res;  
}
```

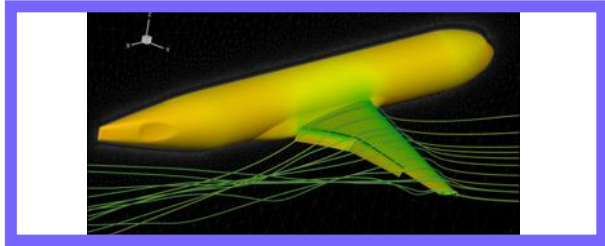
Instead of Chapel code, call an external procedure

```
var myLocalArr: [1..10] int;  
increment(myLocalArr);
```

```
use BlockDist;  
var myDistributedArr = blockDist.createArray({1..10}, int);  
increment(myDistributedArr);
```

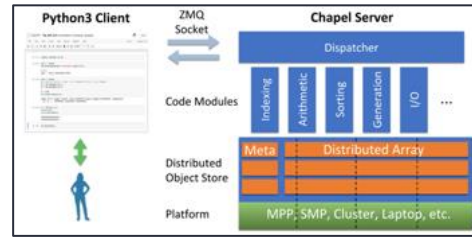
# Applications

# Applications of Chapel



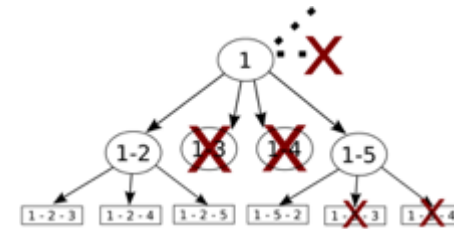
## CHAMPS: 3D Unstructured CFD

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



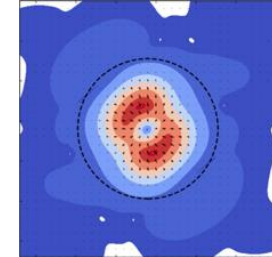
## Arkouda: Interactive Data Science at Massive Scale

Mike Merrill, Bill Reus, et al.  
U.S. DoD



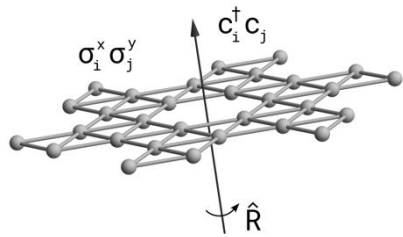
## ChOp: Chapel-based Optimization

T. Carneiro, G. Helbecque, N. Melab, et al.  
INRIA, IMEC, et al.



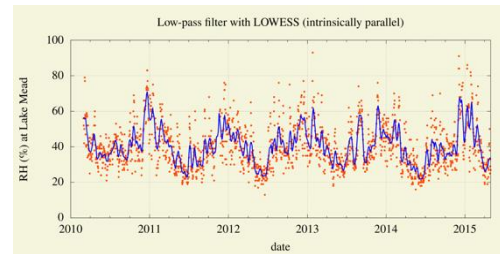
## ChpUltra: Simulating Ultralight Dark Matter

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



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

Tom Westerhout  
Radboud University



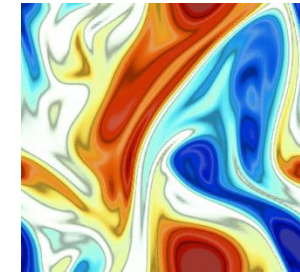
## Desk dot chpl: Utilities for Environmental Eng.

Nelson Luis Dias  
The Federal University of Paraná, Brazil



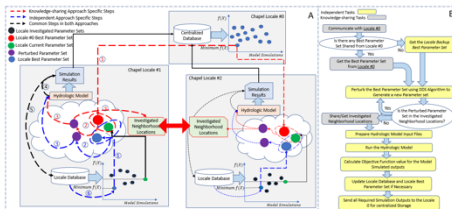
## RapidQ: Mapping Coral Biodiversity

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



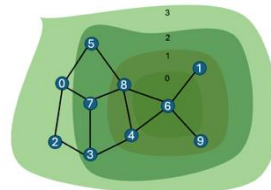
## ChapQG: Layered Quasigeostrophic CFD

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



## Chapel-based Hydrological Model Calibration

Marjan Asgari et al.  
University of Guelph



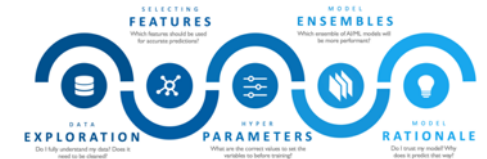
## Arachne Graph Analytics

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



## Modeling Ocean Carbon Dioxide Removal

Scott Bachman Brandon Neth, et al.  
[C]Worthy



## Cray AI HyperParameter Optimization (HPO)

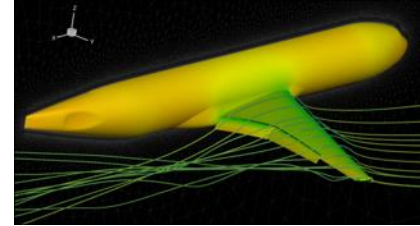
Ben Albrecht et al.  
Cray Inc. / HPE

[images provided by their respective teams and used with permission]

# Applications of Chapel: CHAMPS

## What is it?

- 3D unstructured CFD framework for airplane simulation
- ~100+k lines of Chapel written since 2019



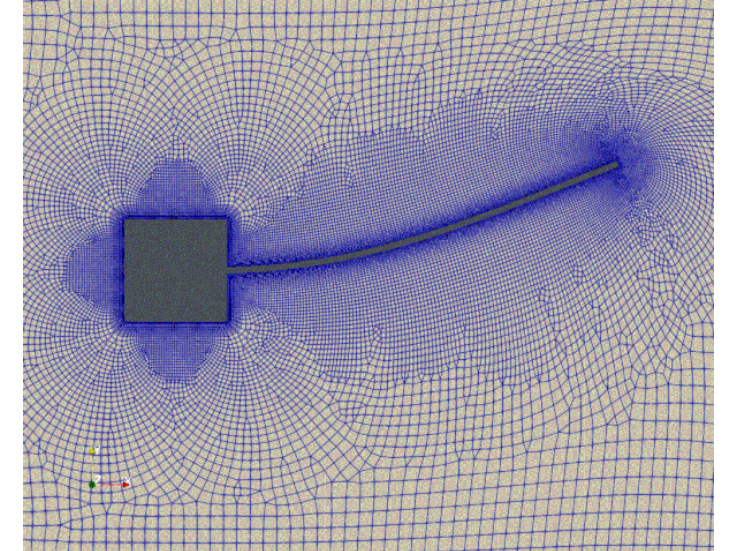
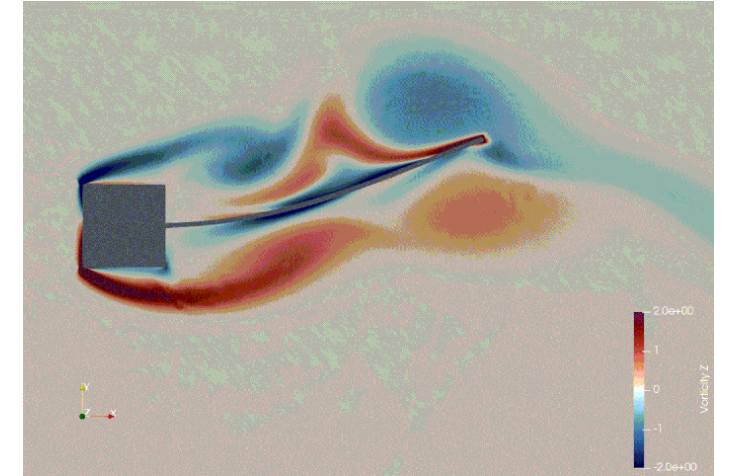
## Who wrote it?

- Professor Éric Laurendeau's students + postdocs at Polytechnique Montreal



## Why Chapel?

- performance and scalability competitive with MPI + C++
- students found it far more productive to use
- enabled them to compete with more established CFD centers



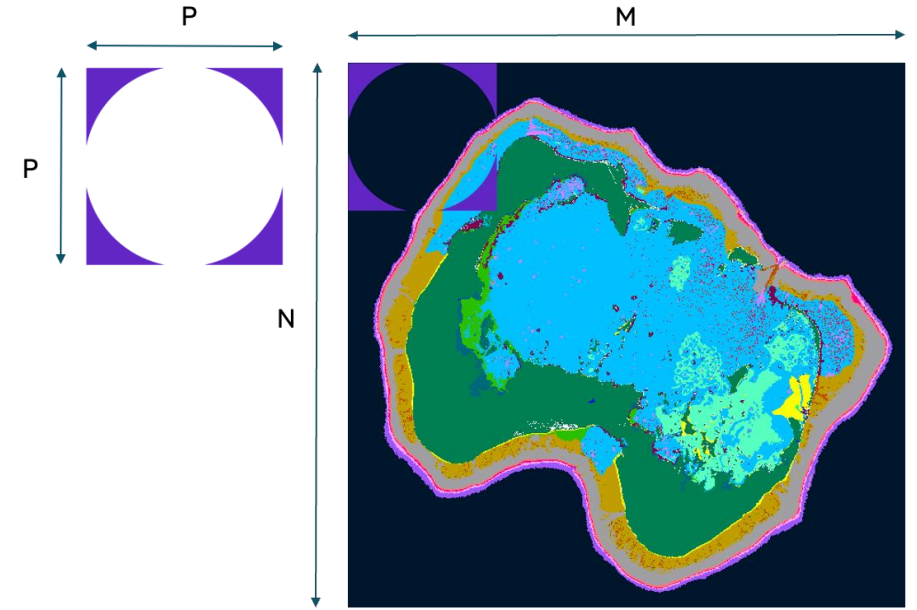
# Applications of Chapel: Biodiversity

## What is it?

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

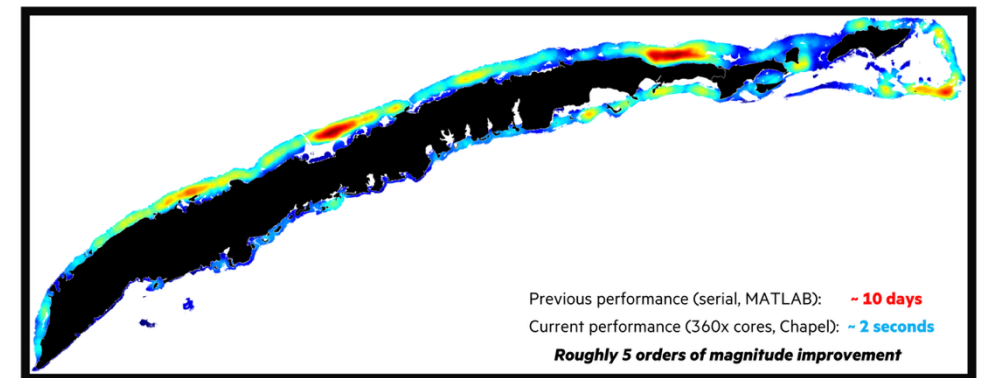
## Who wrote it?

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



## Why Chapel?

- easy transition from Matlab, which they had been using
- massive performance improvement:
  - previous ~10-day run finished in ~2 seconds using 360 cores
- enabled unanticipated algorithmic improvements
  - from  $O(M \cdot N \cdot P)$  habitat diversity to  $O(M \cdot N \cdot P^3)$  spectral diversity
  - Added another ~90 lines of code to make it GPU-enabled
  - ~4-week desktop run  $\rightarrow$  ~20 minutes on 20 nodes / 512 GPUs



# Why Chapel?

## **Get out of the way of science**

- Let scientists express their computations succinctly
- Python and other languages are already good at this
  - But to get performance and scalability, you tend to give up the nice abstractions

## **Easier to write, easier to read, easier to maintain**

- Lowers the barrier to entry for new contributors
- You don't need to be a MPI savant to get good distributed performance
- You don't need to be a CUDA wizard to use GPUs efficiently

# Applications: Arkouda

# Data Science In Python at scale?

**Motivation:** Imagine you work with...

- ...Python-based data scientists
- ...HPC-scale data science problems to solve
- ...access to HPC systems



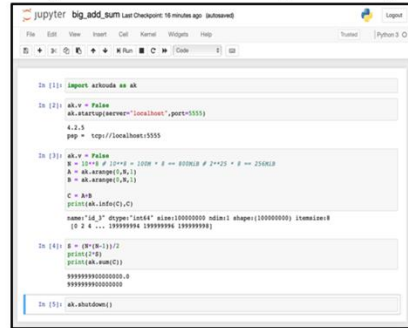
How will you leverage your Python programmers to get your work done?

# What is Arkouda?

**Q:** “What is Arkouda?”



**Arkouda Client**  
(written in Python)

A screenshot of a Jupyter Notebook interface. The code in the notebook shows the import of the arkouda module as 'ak', followed by several lines of code that create arrays, perform operations, and print results. The output of the code is visible in the cells below the code.

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

In [2]: ak.N = 1000000
ak.startup(server="localhost", port=5555)
4.2.5
pqp = http://localhost:5555

In [3]: ak.N = 1000000
N = 1000000 # 2**19 = 524288 * 2 = 1048576
A = ak.arange(0, N-1)
B = ak.arange(0, N-1)
C = A+B
print(ak.info(C))

name: 'ak' dtype: 'uint64' size: 1000000000 bytes shape: (1000000000,) (min: 0, max: 1999999999)

In [4]: B = B*(B-1)/2
print(B[0])
print(ak.sum(C))
999999999999999999.0
999999999999999999.0

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



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

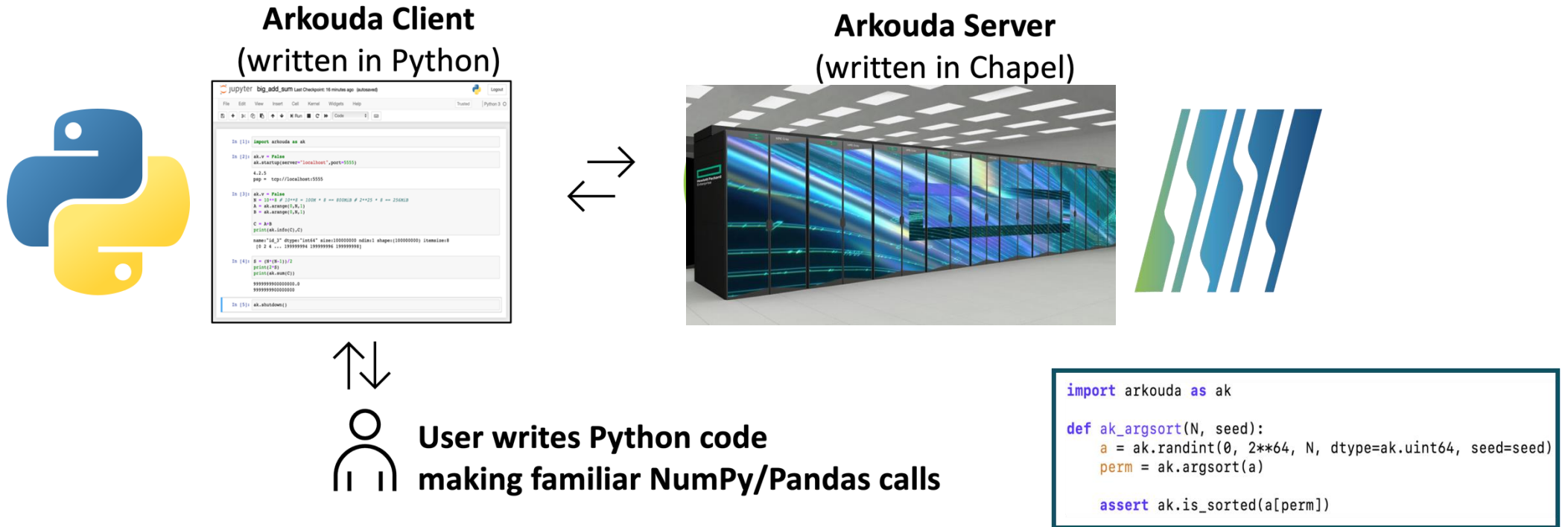
```
import arkouda as ak

def ak_argsort(N, seed):
    a = ak.randint(0, 2**64, N, dtype=ak.uint64, seed=seed)
    perm = ak.argsort(a)

    assert ak.is_sorted(a[perm])
```

# What is Arkouda?

**Q:** “What is Arkouda?”

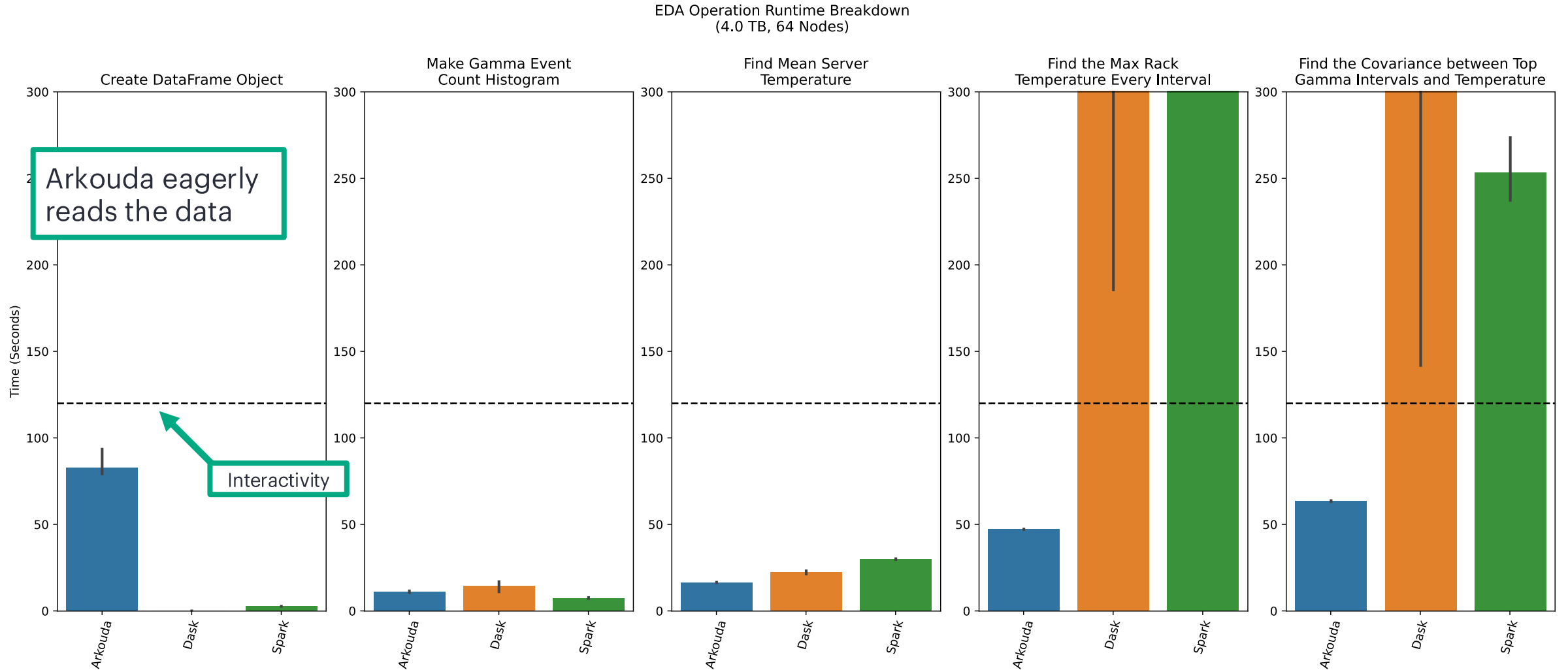


**A1:** “A scalable version of NumPy / Pandas routines for data scientists”

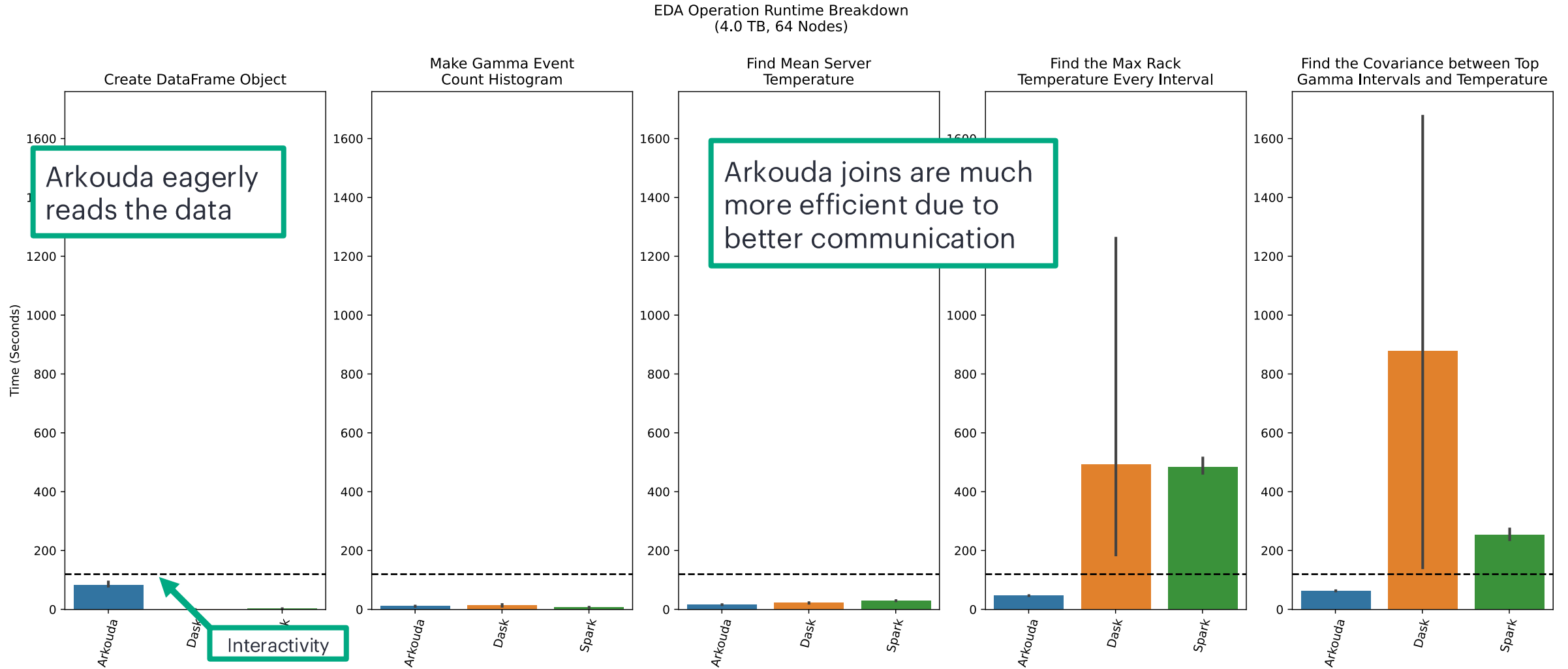
**A2:** “A framework for driving supercomputers interactively from Python”

# Chapel/Arkouda Performance

# Arkouda/Dask/Spark Comparison: 64 nodes w/ 4 TB



# Arkouda/Dask/Spark Comparison: Zoomed out



# Summary

## Chapel is built from the ground up for productive parallel computing at any scale

- Language features express clear and concise computation

## Chapel applications can meet or beat low-level approaches while still...

- ...avoiding unnecessary boilerplate
- ...being human-readable
- ...working with existing workflows



### 7 Questions for Bill Reus: Interactive Supercomputing with Chapel for Cybersecurity

Posted on February 12, 2025.

Tags: User Experiences Interviews Data Analysis Arkouda

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

Part of a series: [7 Questions for Chapel Users](#)

<https://chapel-lang.org/blog/posts/7qs-reus>

*" I was on the verge of resigning myself to learning MPI when I first encountered Chapel. After writing my first Chapel program, I knew I had found something much more appealing."*

# Ways to engage with the Chapel Community

## Synchronous Community Events

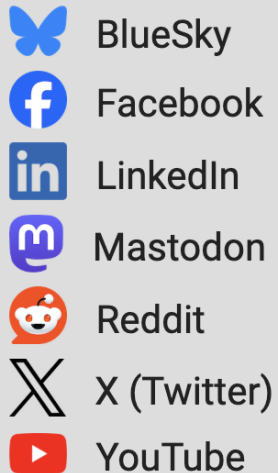
- [Project Meetings](#), weekly
- [Deep Dive / Demo Sessions](#), weekly timeslot
- [Chapel Teaching Meet-up](#), monthly
- [ChapelCon](#) (formerly CHI UW), annually

## Asynchronous Communications

- [Chapel Blog](#), typically ~2 articles per month
- [Community Newsletter](#), quarterly
- [Announcement Emails](#), around big events

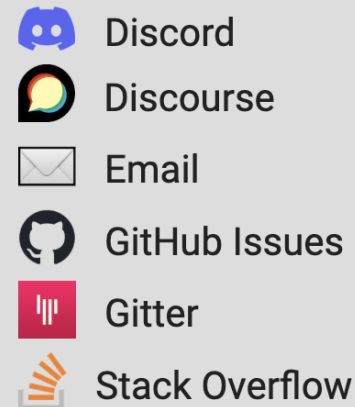
## Social Media

### FOLLOW US



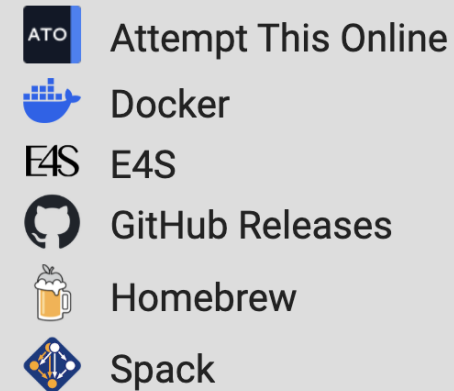
## Discussion Forums

### GET IN TOUCH



## Ways to Use Chapel

### GET STARTED



(from the footer of [chapel-lang.org](https://chapel-lang.org))