



Productive Programming in Chapel: A Computation-Driven Introduction

Short Introduction to Task Parallelism

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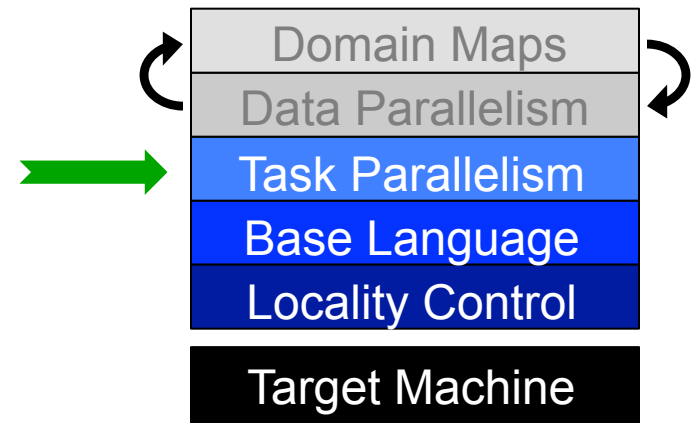


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Outline

- ✓ Motivation
- ✓ Chapel Background and Themes
- ✓ Learning the Base Language with n-body
- Short Introduction to Task Parallelism
- Hands-On 1: Hello World
- Short Introduction to Locality
- Data Parallelism with Jacobi
- Hands-On 2: Mandelbrot
- Project Status, Next Steps





Defining our Terms

Task: a unit of computation that can/should execute in parallel with other tasks

Thread: a system resource that executes tasks

- not exposed in the language
- occasionally exposed in the implementation

Task Parallelism: a style of parallel programming in which parallelism is driven by programmer-specified tasks

(in contrast with):

Data Parallelism: a style of parallel programming in which parallelism is driven by computations over collections of data elements or their indices

Task Parallelism: Begin Statements

```
// create a fire-and-forget task for a statement
begin writeln("hello world");
writeln("goodbye");
```

Possible outputs:

```
hello world
goodbye
```

```
goodbye
hello world
```

Cobegins/Serial by Example: QuickSort

```

proc quickSort(arr: [?D],
               thresh = log2(here.maxTaskPar),
               depth = 0,
               low: int = D.low,
               high: int = D.high) {
  if high - low < 8 {
    bubbleSort(arr, low, high);
  } else {
    const pivotVal = findPivot(arr, low, high);
    const pivotLoc = partition(arr, low, high, pivotVal);
    serial (depth >= thresh) do cobegin {
      quickSort(arr, thresh, depth+1, low, pivotLoc-1);
      quickSort(arr, thresh, depth+1, pivotLoc+1, high);
    }
  }
}

```



Cobegins/Serial by Example: QuickSort

```
proc quickSort(arr: [?D],  
               depth = 0,  
               low: int = D.low,  
               high: int = D.high) {  
  if high - low < 8 {  
    bubbleSort(arr, low, high);  
  } else {  
    const pivotVal = findPivot(arr, low, high);  
    const pivotLoc = partition(arr, low, high, pivotVal);  
    serial (here.runningTasks > here.maxTaskPar) do  
      cobegin {  
        quickSort(arr, depth+1, low, pivotLoc-1);  
        quickSort(arr, depth+1, pivotLoc+1, high);  
      }  
  }  
}
```



Task Parallelism: Cobegin Statements

```
// create a task per child statement  
cobegin {  
    producer(1);  
    producer(2);  
    consumer(1);  
} // implicit join of the three tasks here
```




Task Parallelism: Coforall Loops

```
// create a task per iteration
coforall t in 0..#numTasks {
    writeln("Hello from task ", t, " of ", numTasks);
} // implicit join of the numTasks tasks here

writeln("All tasks done");
```

Sample output:

```
Hello from task 2 of 4
Hello from task 0 of 4
Hello from task 3 of 4
Hello from task 1 of 4
All tasks done
```

Comparison of Begin, Cobegin, and Coforall

begin:

- Use to create a dynamic task with an unstructured lifetime
- “fire and forget”

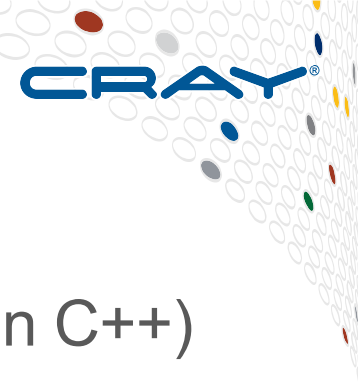
cobegin:

- Use to create a related set of heterogeneous tasks
- ...or a small, finite set of homogenous tasks
- The parent task depends on the completion of the tasks

coforall:

- Use to create a fixed or dynamic # of homogenous tasks
- The parent task depends on the completion of the tasks

Note: All these concepts can be composed arbitrarily



Task Parallelism: Data-Driven Synchronization

- 1) ***atomic variables***: support atomic operations (as in C++)
 - e.g., compare-and-swap; atomic sum, etc.
- 2) ***single-assignment variables***: reads block until assigned
- 3) ***synchronization variables***: store full/empty state
 - by default, reads/writes block until the state is full/empty



Bounded Buffer Producer/Consumer Example

```
begin producer();  
consumer();  
  
// 'sync' types store full/empty state along with value  
var buff$: [0..#buffersize] sync real;  
  
proc producer() {  
  var i = 0;  
  for ... {  
    i = (i+1) % buffersize;  
    buff$[i] = ...; // writes block until empty, leave full  
  } }  
  
proc consumer() {  
  var i = 0;  
  while ... {  
    i = (i+1) % buffersize;  
    ...buff$[i]...; // reads block until full, leave empty  
  } }
```

Synchronization Variables

• Syntax

```
sync-type:
  sync type
```

• Semantics

- Stores *full/empty* state along with normal value
- Defaults to *full* if initialized, *empty* otherwise
- Default read blocks until *full*, leaves *empty*
- Default write blocks until *empty*, leaves *full*

• Examples: Critical sections and futures

```
var future$: sync real;

begin future$ = compute();
res = computeSomethingElse();
useComputedResults(future$, res);
```

```
var lock$: sync bool;

lock$ = true;
critical();
var lockval = lock$;
```

Atomic Variables

- **Syntax**

```
sync-type:
  atomic type
```

- **Semantics**

- Supports operations on variable atomically w.r.t. other tasks
- Based on C/C++ atomic operations

- **Example: Trivial barrier**

```
var count: atomic int, done: atomic bool;

proc barrier(numTasks) {
  const myCount = count.fetchAdd(1);
  if (myCount < numTasks - 1) then
    done.waitFor(true);
  else
    done.testAndSet();
}
```

Atomic Methods

- `read() : t` return current value
- `write(v : t)` store *v* as current value
- `exchange(v : t) : t` store *v*, returning previous value
- `compareExchange(old : t, new : t) : bool`
 store *new* iff previous value was *old*; returns true on success
- `waitFor(v : t)` wait until the stored value is *v*
- `add(v : t)` add *v* to the value atomically
- `fetchAdd(v : t)` same, returning pre-sum value
 (*sub, or, and, xor* also supported similarly)
- `testAndSet()` like *exchange(true)* for atomic
 bool
- `clear()` like *write(false)* for atomic bool

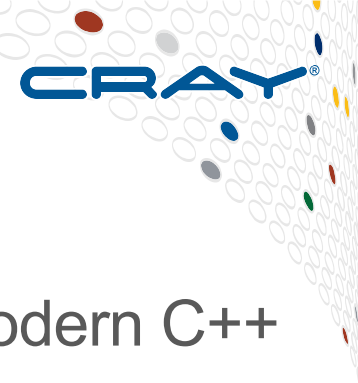
Comparison of Synchronization Types

sync/single:

- Best for producer/consumer style synchronization
 - “this task should block until something happens”
- Use single for write-once values

atomic:

- Best for uncoordinated accesses to shared state



Other Task Parallel Concepts

- **atomic variables:** support atomics ops, similar to modern C++
- **sync/single variables:** support producer/consumer patterns
- **sync statements:** join unstructured tasks
- **serial statements:** conditionally squash parallelism



Other Task Parallel Features

Current:

- **serial statements to conditionally squash parallelism**
- **sync statements to join dynamically generated tasks**

Planned:

- **task-private variables**
- **task teams to support**
 - collective operations (barriers, joins, reductions, etc.)
 - thread scheduling policies

Smith-Waterman Algorithm for Sequence Alignment





Smith-Waterman

Goal: Determine the similarities/differences between two protein sequences/nucleotides.

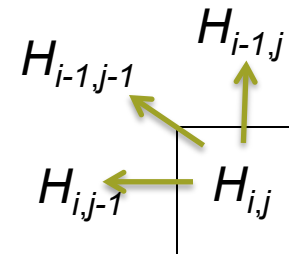
- e.g., ACACACTA and AGCACACA*

Basis of Computation: Defined via a recursive formula:

$$H(i,0) = 0$$

$$H(0,j) = 0$$

$$H(i,j) = f(H(i-1,j-1), H(i-1,j), H(i,j-1))$$



Caveat: *This is a classic, rather than cutting-edge sequence alignment algorithm, but it illustrates an important parallel paradigm: wavefront computation*

*Source of running example: Wikipedia



Smith-Waterman

Naïve Task-Parallel Approach:

```
proc computeH(i, j) {  
  if (i == 0 || j == 0) then  
    return 0;  
  else  
    var h_NW, h_N, h_W: int;  
  
    cobegin {  
      h_NW = computeH(i-1, j-1);  
      h_N  = computeH(i-1, j);  
      h_W  = computeH(i, j-1);  
    }  
  
    return f(h_NW, h_N, h_W);  
}
```

Note: Recomputes most subexpressions redundantly

This is a case for dynamic programming!

Dynamic Programming Approach:

	0	1	2	3	4	5	6	7	8
0	0	0	0	0	0	0	0	0	0
1	0								
2	0								
3	0								
4	0								
5	0								
6	0								
7	0								
8	0								

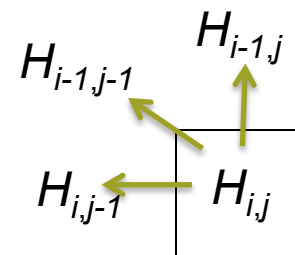
Step 1: Initialize
boundaries to 0

Smith-Waterman

Dynamic Programming Approach:

	0	1	2	3	4	5	6	7	8
0	0	0	0	0	0	0	0	0	0
1	0								
2	0								
3	0								
4	0				etc.				
5	0								
6	0								
7	0								
8	0								

Step 2: Compute cells when we're able to



Smith-Waterman

Dynamic Programming Approach:

	0	1	2	3	4	5	6	7	8
0	0	0	0	0	0	0	0	0	0
1	0	2	1	2	1	2	1	0	2
2	0	1	1	1	1	1	1	0	1
3	0	0	3	2	3	2	3	2	1
4	0	2	2	5	4	5	4	3	4
5	0	1	4	4	7	6	7	6	5
6	0	2	3	6	6	9	8	7	8
7	0	1	4	5	8	8	11	10	9
8	0	2	3	6	7	10	10	10	12

Step 3: Follow trail of breadcrumbs back

Smith-Waterman

Dynamic Programming Approach:

	0	1	2	3	4	5	6	7	8
0	0	0	0	0	0	0	0	0	0
1	0	2	1	2	1	2	1	0	2
2	0	1	1	1	1	1	1	0	1
3	0	0	3	2	3	2	3	2	1
4	0	2	2	5	4	5	4	3	4
5	0	1	4	4	7	6	7	6	5
6	0	2	3	6	6	9	8	7	8
7	0	1	4	5	8	8	11	10	9
8	0	2	3	6	7	10	10	10	12

Step 3: Follow trail of breadcrumbs back

Smith-Waterman

Dynamic Programming Approach:

Step 4: Interpret the path against the original sequences

		A	C	A	C	A	C	T	A	
A G C A C A C A		0	0	0	0	0	0	0	0	
	A	0	2	1	2	1	2	1	0	2
	G	0	1	1	1	1	1	1	0	1
	C	0	0	3	2	3	2	3	2	1
	A	0	2	2	5	4	5	4	3	4
	C	0	1	4	4	7	6	7	6	5
	A	0	2	3	6	6	9	8	7	8
	C	0	1	4	5	8	8	11	10	9
	A	0	2	3	6	7	10	10	10	12

AGCACAC-A
A-CACACTA

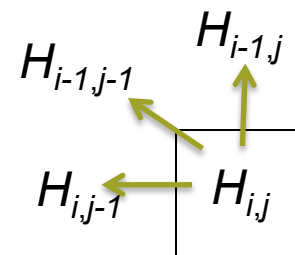
Smith-Waterman

Dynamic Programming Approach:

	0	1	2	3	4	5	6	7	8
0	0	0	0	0	0	0	0	0	0
1	0								
2	0								
3	0								
4	0				etc.				
5	0								
6	0								
7	0								
8	0								

Step 2: Compute cells when we're able to

How could we do this in parallel?



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Smith-Waterman

Data-Parallel Approach:

```

proc computeH(H: [0..n, 0..n] int) {
  for upperDiag in 1..n do
    forall diagPos in 0..#upperDiag {
      const (i,j) = (diagPos+1, upperDiag-diagPos);
      H[i,j] = f(H[i-1,j-1], H[i-1,j], H[i,j-1]);
    }
  for lowerDiag in 1..n-1 do
    forall diagPos in lowerDiag..n-1 by -1 {
      const (i,j) = (diagPos+1, lowerDiag+diagPos);
      H[i,j] = f(H[i-1,j-1], H[i-1,j], H[i,j-1]);
    }
  }
}

```

Loop over upper diagonals serially

Process each diagonal in parallel

Repeat for lower diagonals

Advantages:

- Reasonably clean
(if I got my indexing correct)

Disadvantages:

- Not so great in terms of cache use
- A bit fine-grained
 - small number of iterations per task



Naïve Data-Driven Task-Parallel Approach:

```
proc computeH(H: [0..n, 0..n] int) {  
  const ProbSpace = H.domain.translate(1,1);  
  var NeighborsDone: [ProbSpace] atomic int;  
  var Ready$: [ProbSpace] sync int;
```

Create a domain describing shifted version of H's domain

Arrays to count how many of our 3 neighbors are done; and to signal when we can compute

```
  NeighborsDone[1, ..].add(1);  
  NeighborsDone[.., 1].add(1);  
  NeighborsDone[1, 1].add(1);  
  Ready$[1,1] = 1;
```

Set up boundaries: north and west elements have a neighbor done; top-left is ready

```
  forall (i,j) in ProbSpace {  
    const goNow = Ready$[i,j];  
    H[i,j] = f(H[i-1,j-1], H[i-1,j], H[i,j-1]);  
    const eastReady = NeighborsDone[i, j+1].fetchAdd(1);  
    const seReady = NeighborsDone[i+1,j+1].fetchAdd(1);  
    const southReady = NeighborsDone[i+1,j].fetchAdd(1);  
    if (eastReady == 2) then Ready$[i, j+1] = 1;  
    if (seReady == 2) then Ready$[i+1,j+1] = 1;  
    if (southReady == 2) then Ready$[i+1,j] = 1;  
  }  
}
```

Create a task per matrix element and have it block until ready

Compute our element

Increment our neighbors' counts

Signal our neighbors as ready if we're the third



Naïve Data-Driven Task-Parallel Approach:

```
proc computeH(H: [0..n, 0..n] int) {  
  const ProbSpace = H.domain.translate(1,1);  
  var NeighborsDone: [ProbSpace] atomic int;  
  var Ready$: [ProbSpace] sync int;
```

```
  NeighborsDone[1, ..].add(1);  
  NeighborsDone[.., 1].add(1);  
  NeighborsDone[1, 1].add(1);  
  Ready$[1,1] = 1;
```

```
  forall (i,j) in ProbSpace {  
    const goNow = Ready$[i,j];  
    H[i,j] = f(H[i-1,j-1], H[i-1,j], H[i,j-1]);  
    const eastReady = NeighborsDone[i, j+1].fetchAdd(1);  
    const seReady = NeighborsDone[i+1,j+1].fetchAdd(1);  
    const southReady = NeighborsDone[i+1,j ].fetchAdd(1);  
    if (eastReady == 2) then Ready$[i, j+1] = 1;  
    if (seReady == 2) then Ready$[i+1,j+1] = 1;  
    if (southReady == 2) then Ready$[i+1,j ] = 1;  
  }  
}
```

Disadvantages:

- Still not great in cache use
- Uses n^2 tasks
- Most spend most of their time blocking



Slightly Less Naïve Data-Driven Task-Parallel Approach:

```
proc computeH(H: [0..n, 0..n] int) {  
  const ProbSpace = H.domain.translate(1,1);  
  var NeighborsDone: [ProbSpace] atomic int;
```

```
  NeighborsDone[1, ..].add(1);  
  NeighborsDone[.., 1].add(1);  
  NeighborsDone[1, 1].add(1);  
  sync { computeHHelp(1,1); }
```

sync to ensure they're all done before we go on

```
proc computeHHelp(i,j) {  
  H[i,j] = f(H[i-1,j-1], H[i-1,j], H[i,j-1]);  
  const eastReady = NeighborsDone[i, j+1].fetchAdd(1);  
  const seReady = NeighborsDone[i+1,j+1].fetchAdd(1);  
  const southReady = NeighborsDone[i+1,j ].fetchAdd(1);  
  if (eastReady == 2) then begin computeHHelp(i, j+1);  
  if (seReady == 2) then begin computeHHelp(i+1,j+1);  
  if (southReady == 2) then begin computeHHelp(i+1,j );  
}
```

Rather than create the tasks *a priori*, fire them off once we know they're ready to compute



Slightly Less Naïve Data-Driven Task-Parallel Approach:

```
proc computeH(H: [0..n, 0..n] int) {  
  const ProbSpace = H.domain.translate(1,1);  
  var NeighborsDone: [ProbSpace] atomic int;
```

```
  NeighborsDone[1, ..].add(1);  
  NeighborsDone[.., 1].add(1);  
  NeighborsDone[1, 1].add(1);  
  sync { computeHHelp(1,1); }
```

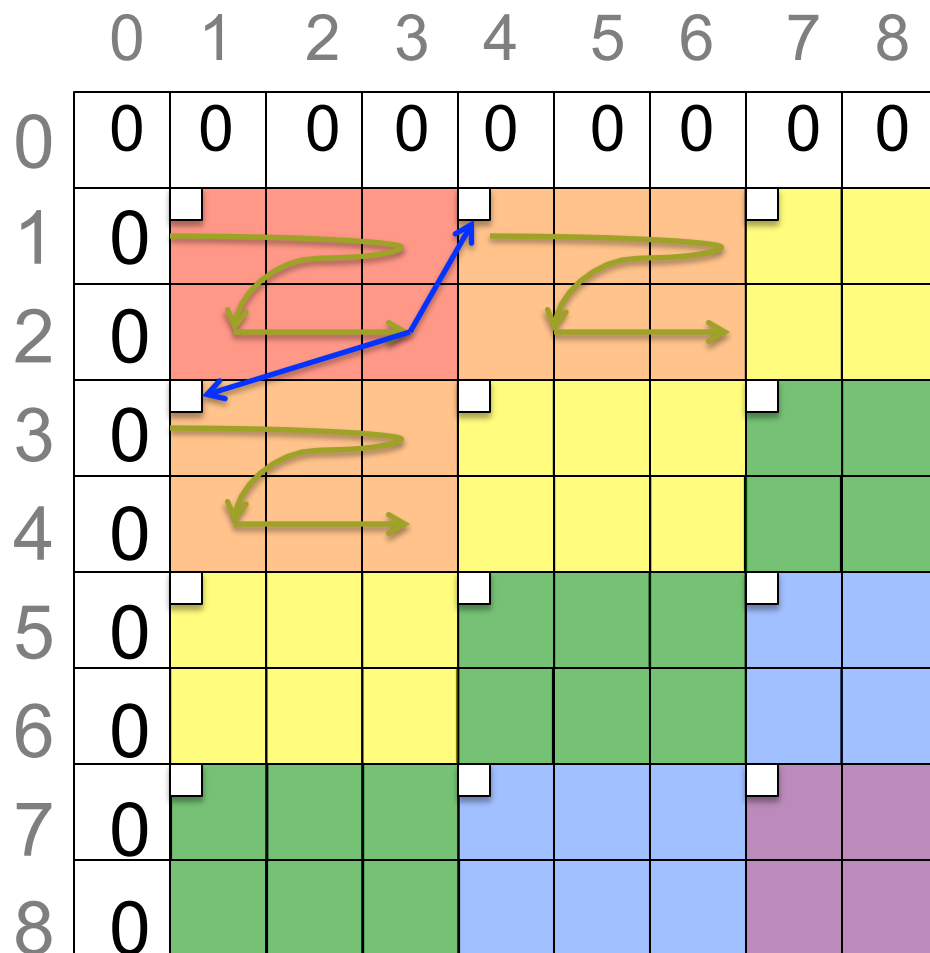
Disadvantages:

- Still uses a lot of tasks
- Each task is very fine-grained

```
proc computeHHelp(i,j) {  
  H[i,j] = f(H[i-1,j-1], H[i-1,j], H[i,j-1]);  
  const eastReady = NeighborsDone[i, j+1].fetchAdd(1);  
  const seReady = NeighborsDone[i+1,j+1].fetchAdd(1);  
  const southReady = NeighborsDone[i+1,j ].fetchAdd(1);  
  if (eastReady == 2) then begin computeHHelp(i, j+1);  
  if (seReady == 2) then begin computeHHelp(i+1,j+1);  
  if (southReady == 2) then begin computeHHelp(i+1,j );  
}
```


Smith-Waterman

Coarsening the Parallelism into Chunks:



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Chunked Data-Driven Task-Parallel Approach:

```
proc computeH(H: [0..n, 0..n] int) {  
  const ProbSpace = H.domain.translate(1,1);  
  const StrProbSpace = ProbSpace by (rowsPerChunk, colsPerChunk);  
  var NeighborsDone: [StrProbSpace] atomic int;
```

Use strided array for atomics

```
  NeighborsDone[1, ..].add(1);  
  NeighborsDone[.., 1].add(1);  
  NeighborsDone[1, 1].add(1);  
  sync { computeHHelp(1,1); }
```

Change helper to iterate over a chunk serially

```
  proc computeHHelp(x,y) {  
    for (i,j) in ProbSpace[x..#rowsPerChunk, y..#colsPerChunk] do  
      H[i,j] = f(H[i-1,j-1], H[i-1,j], H[i,j-1]);  
      const eastReady = NeighborsDone[x, y+colsPerChunk].fetchAdd(1);  
      const seReady = NeighborsDone[x+rowsPerChunk, y+colsPerChunk].fetchAdd(1);  
      const southReady = NeighborsDone[x+rowsPerChunk, y].fetchAdd(1);  
      if (eastReady == 2) then begin computeHHelp(x, y+colsPerChunk);  
      if (seReady == 2) then begin computeHHelp(x+rowsPerChunk, y+colsPerChunk);  
      if (southReady == 2) then begin computeHHelp(x+rowsPerChunk, y);  
    }  
  }
```

Stride indices to get to next chunk's origin

Questions about Task Parallelism in Chapel?





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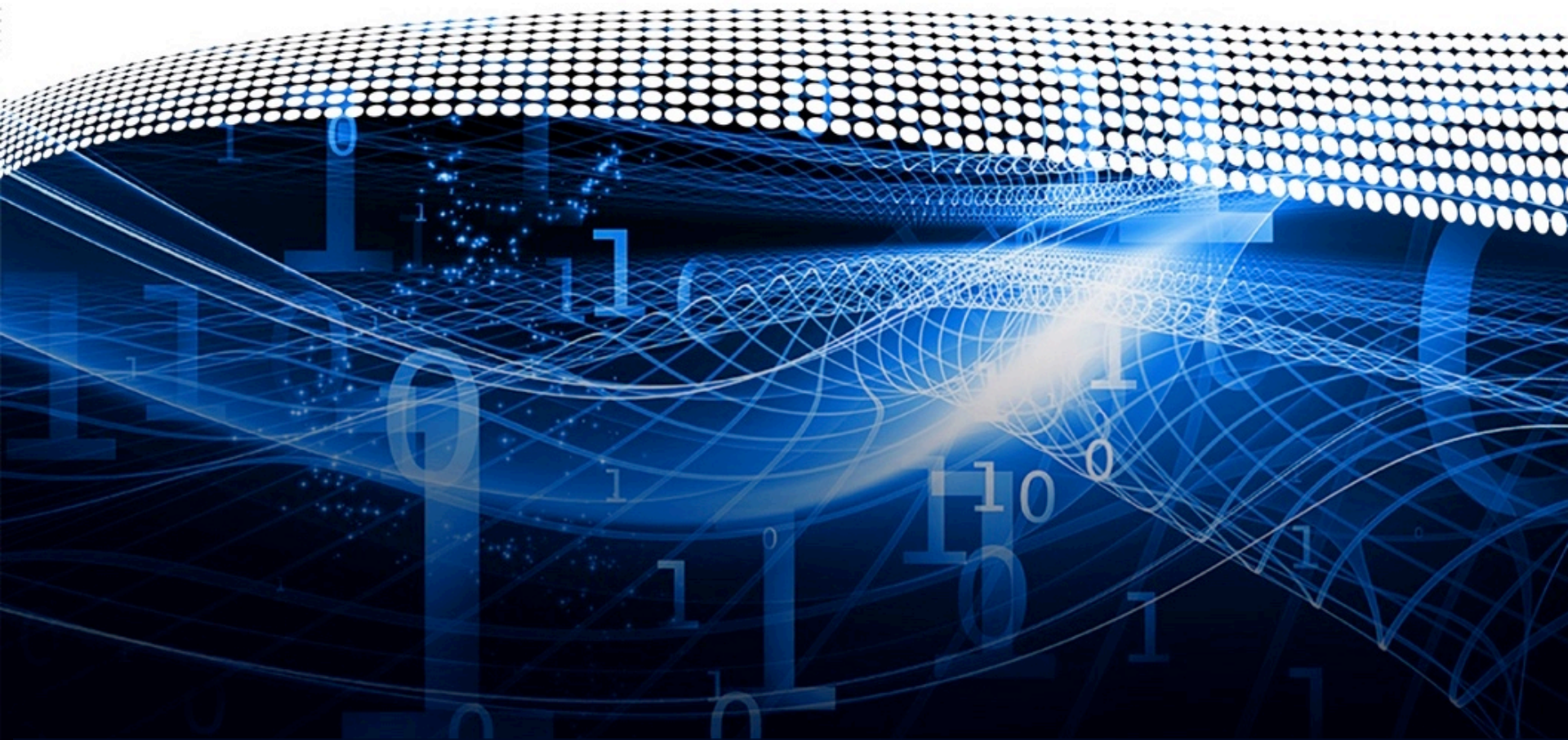
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<http://github.com/chapel-lang/chapel/>