

# Data Parallelism



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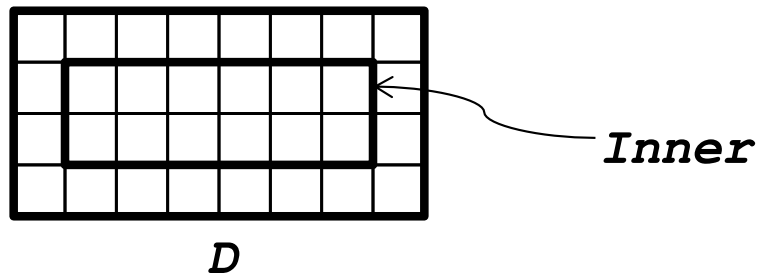


# Domains

## *Domain:*

- A first-class index set
- The fundamental Chapel concept for data parallelism

```
config const m = 4, n = 8;  
  
const D = {1..m, 1..n};  
const Inner = {2..m-1, 2..n-1};
```

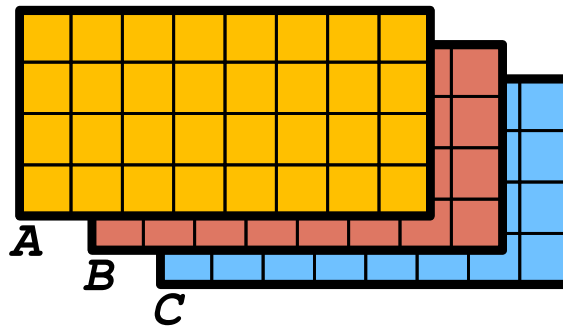


# Domains

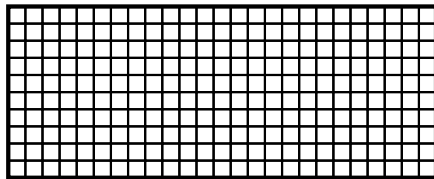
## Domain:

- A first-class index set
- The fundamental Chapel concept for data parallelism
- Useful for declaring arrays and computing with them

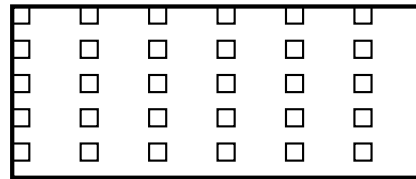
```
config const m = 4, n = 8;  
  
const D = {1..m, 1..n};  
const Inner = {2..m-1, 2..n-1};  
  
var A, B, C: [D] real;
```



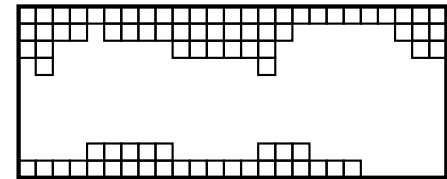
# Chapel Domain Types



*dense*



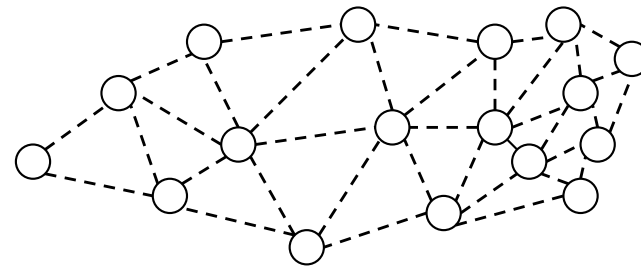
*strided*



*sparse*

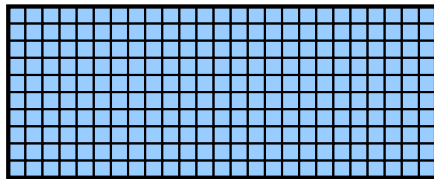


*associative*

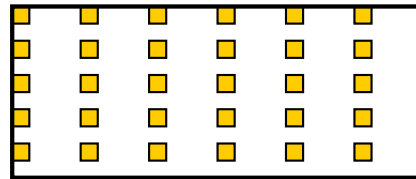


*unstructured*

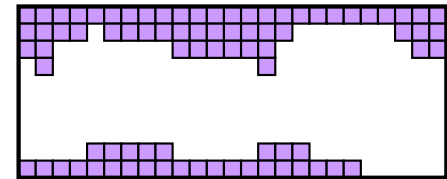
# Chapel Array Types



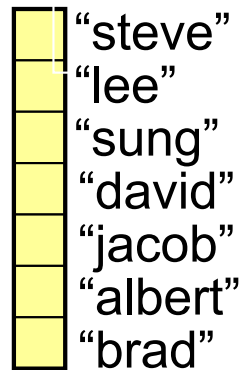
*dense*



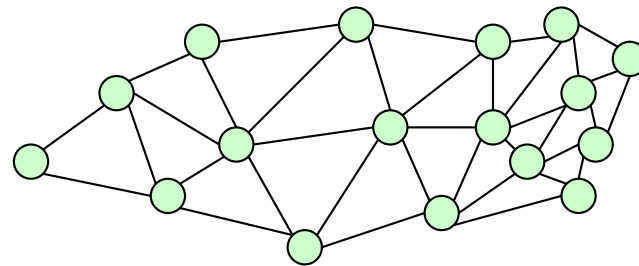
*strided*



*sparse*



*associative*



*unstructured*

# Data Parallelism By Example: STREAM Triad

```
const ProblemSpace = {1..m};
```



```
var A, B, C: [ProblemSpace] real;
```



```
forall (a,b,c) in zip(A,B,C) do  
  a = b + alpha*c;
```

# Forall Loops

## Forall loops: Central concept for data parallel computation

- Like for-loops, but parallel
- Implementation details determined by iterand (e.g.,  $D$  below)
  - specifies number of tasks, which tasks run which iterations, ...
  - in practice, typically uses a number of tasks appropriate for target HW

```
forall (i,j) in D do  
    A[i,j] = i + j/10.0;
```

1.1	1.2	1.3	1.4	1.5	1.6	1.7	1.8
2.1	2.2	2.3	2.4	2.5	2.6	2.7	2.8
3.1	3.2	3.3	3.4	3.5	3.6	3.7	3.8
4.1	4.2	4.3	4.4	4.5	4.6	4.7	4.8

- **Forall loops assert...**
  - ...parallel safety: OK to execute iterations simultaneously
  - ...order independence: iterations could occur in any order
  - ...serializability: all iterations could be executed by one task
    - e.g., can't have synchronization dependences between iterations





# Comparison of Loops: For, Forall, and Coforall



## For loops: executed using one task

- use when a loop must be executed serially
- or when one task is sufficient for performance

## Forall loops: typically executed using $1 < \#tasks \ll \#iters$

- use when a loop *should* be executed in parallel...
- ...but *can* legally be executed serially
- use when desired  $\# \text{ tasks} \ll \# \text{ of iterations}$

## Coforall loops: executed using a task per iteration

- use when the loop iterations *must* be executed in parallel
- use when you want  $\# \text{ tasks} == \# \text{ of iterations}$
- use when each iteration has substantial work



# Forall Intents

- **Tell how to “pass” variables from outer scopes to tasks**
  - Similar to argument intents in syntax and philosophy
    - also adds a “reduce intent”, similar to OpenMP
  - Design principles:
    - “principle of least surprise”
    - avoid simple race conditions
    - avoid copies of (potentially) expensive data structures



# forall Intent Examples: Scalars

```

var sum: real;
forall i in 1..n do                                // default intent of scalars is 'const in'
    sum += computeMyResult(i);                       // so this is illegal (and avoids a race)

var sum: real;
forall i in 1..n with (ref sum) do                 // override default intent
    sum += computeMyResult(i);                       // we've now requested a race

var sum: real;
forall i in 1..n with (+ reduce sum) do             // override default intent
    sum += computeMyResult(i);                       // each task accumulates into its own copy
// on loop exit, all tasks combine their results into original 'sum'

```

# forall Intent Examples: Arrays

```

var sum: [1..1000] real;
forall i in 1..1000 do
    sum[i] = computeMyResult(i);

```

*// default intent for arrays is 'ref'*  
*// (avoids array copies by default)*

```

var sum: [1..1000] real;
forall i in 1..1000 with (in sum) do
    sum[i] = computeMyResult(i);

```

*// override default intent: "copy in"*  
*// each task has its own copy now*

```

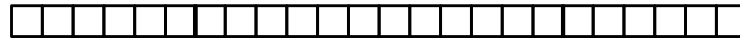
var sum: [1..1000] real;
forall i in 1..n with (+ reduce sum) do
    sum[computeBucket(i)] += 1;

```

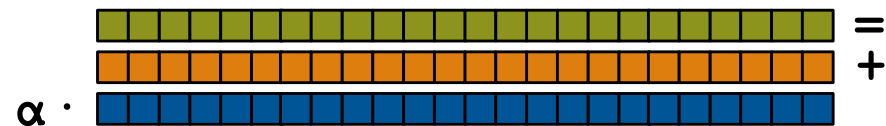
*// request reduce on exit*  
*// each task has its own copy now*  
*// on loop exit, tasks combine their results into original 'sum', computing a histogram*

# Data Parallelism By Example: STREAM Triad

```
const ProblemSpace = {1..m};
```



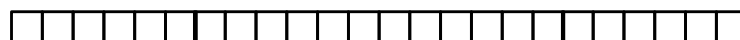
```
var A, B, C: [ProblemSpace] real;
```



```
forall (a,b,c) in zip(A,B,C) do  
  a = b + alpha*c;
```

# Data Parallelism By Example: STREAM Triad

```
const ProblemSpace = {1..m};
```



```
var A, B, C: [ProblemSpace] real;
```



`A = B + alpha * C;`    *// equivalent to the previous zippered forall version*

# Function promotion

- **Scalar functions may be called with array arguments**
  - functions expecting arguments of type  $t$  can be passed array-of- $t$ 
    - results in data parallel invocation of function

```
proc foo(x: int, y: int) {
  return 2*x + y;
}
writeln(foo(3,4));           // prints 10
writeln(foo([1, 2, 4], [2, 3, 4])); // prints 4 7 12
```

- Promotion is equivalent to zippered iteration:

```
foo(A, B);
```

==

```
forall (a,b) in zip(A, B) do
  foo(a, b);
```

- **Ranges/domains can also promote functions:**

```
writeln(foo(1..3, 1..6 by 2)); // prints 3 7 11
```

# Implication of Zippered Promotion Semantics

Whole-array operations are implemented element-wise...

`A = B + alpha * C;`  $\Rightarrow$  `forall (a,b,c) in zip(A,B,C) do  
a = b + alpha * c;`

...rather than operator-wise.

`A = B + alpha * C;`  $\nRightarrow$  `T1 = alpha * C;  
A = B + T1;`



# Implication of Zippered Promotion Semantics

Whole-array operations are implemented element-wise...

$A = B + \text{alpha} * C;$   $\Rightarrow$  `forall (a,b,c) in zip(A,B,C) do  
a = b + alpha * c;`

$\Rightarrow$  No temporary arrays required by semantics

$\Rightarrow$  No surprises in memory requirements

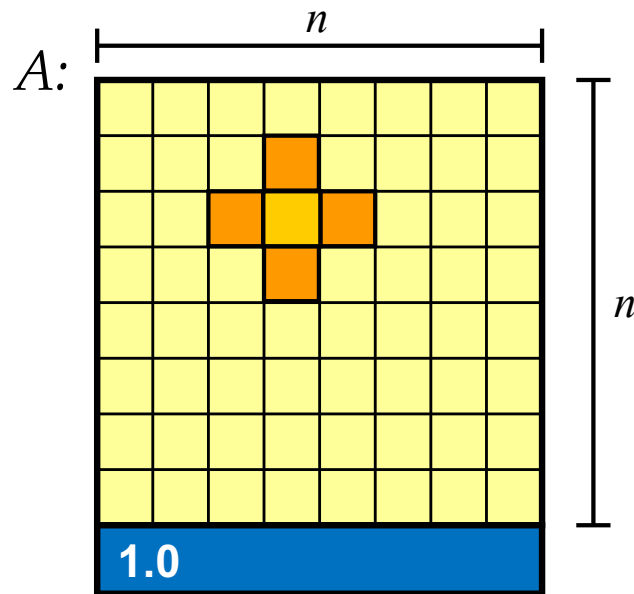
$\Rightarrow$  Friendlier to cache utilization

$\Rightarrow$  Differs from traditional array language semantics

$A[D] = A[D-one] + A[D+one];$   $\Rightarrow$  `forall (a1, a2, a3)  
in (A[D], A[D-one], A[D+one]) do  
a1 = a2 + a3;`

Read/write race!

# Data Parallelism by Example: Jacobi Iteration



repeat until max  
change  $< \epsilon$



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# Jacobi Iteration in Chapel

```

config const n = 6,
               epsilon = 1.0e-5;

const BigD = {0..n+1, 0..n+1},
          D = BigD[1..n, 1..n],
          LastRow = D.exterior(1,0);

var A, Temp : [BigD] real;

A[LastRow] = 1.0;

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

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

writeln(A);

```



# Jacobi Iteration in Chapel

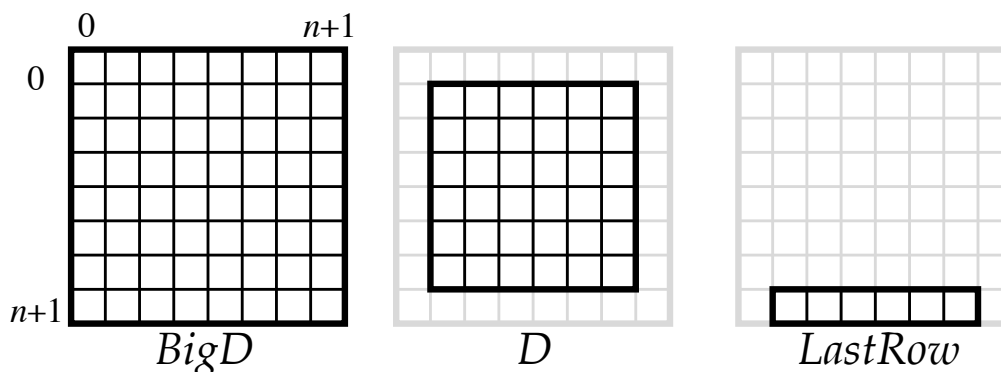
```
config const n = 6,
            epsilon = 1.0e-5;

const BigD = {0..n+1, 0..n+1},
      D = BigD[1..n, 1..n],
      LastRow = D.exterior(1,0);
```

## Declare domains (first class index sets)

$\{lo..hi, lo2..hi2\} \Rightarrow$  2D rectangular domain, with 2-tuple indices

**Dom1[Dom2]**  $\Rightarrow$  computes the intersection of two domains



**.exterior()**  $\Rightarrow$  one of several built-in domain generators

# Jacobi Iteration in Chapel

```
config const n = 6,  
            epsilon = 1.0e-5;
```

```
const BigD = {0..n+1, 0..n+1},  
            D = BigD[1..n, 1..n],  
            LastRow = D.exterior(1,0);
```

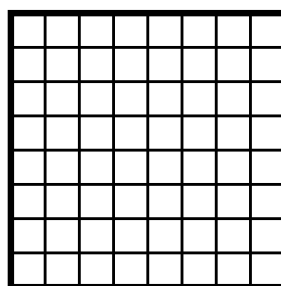
```
var A, Temp : [BigD] real;
```

## Declare arrays

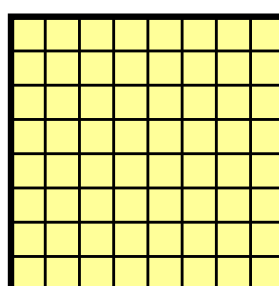
**var**  $\Rightarrow$  can be modified throughout its lifetime

: [**Dom**] **T**  $\Rightarrow$  array of size *Dom* with elements of type *T*

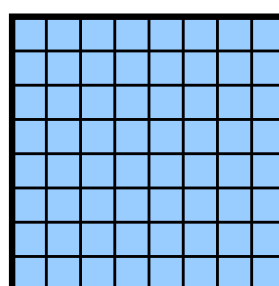
(**no initializer**)  $\Rightarrow$  values initialized to default value (0.0 for reals)



*BigD*



*A*



*Temp*

# Jacobi Iteration in Chapel

```
config const n = 6,
```

## Compute 5-point stencil

**forall** *ind* in *Dom*  $\Rightarrow$  parallel forall expression over *Dom*'s indices,  
binding them to *ind*  
(here, since *Dom* is 2D, we can de-tuple the indices)



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

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

writeln(A);
```

# Jacobi Iteration in Chapel

```
config const n = 6,  
            epsilon = 1.0e-5;
```

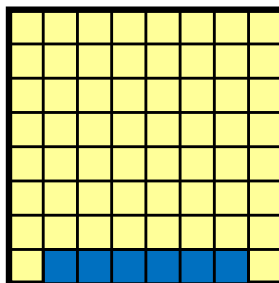
```
const BigD = {0..n+1, 0..n+1},  
            D = BigD[1..n, 1..n],  
            LastRow = D.exterior(1,0);
```

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

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

## Set Explicit Boundary Condition

**Arr[Dom]**  $\Rightarrow$  refer to array slice (“forall i in Dom do ...Arr[i]...”)

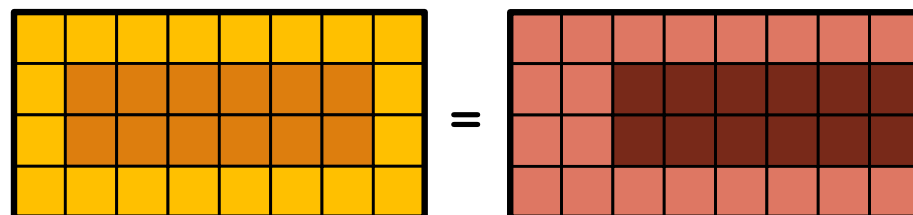


A

# Array Slicing

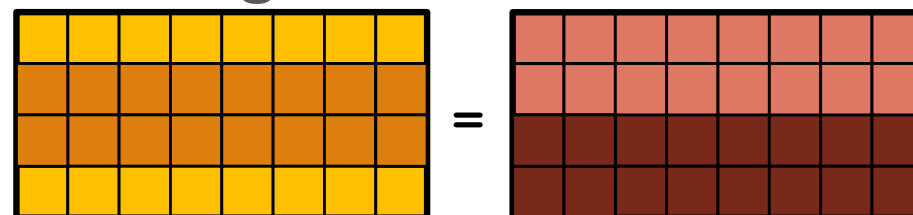
- **Domains can be used to index into arrays**
  - Can be thought of as “promoted array indexing”

```
A[InnerD] = B[InnerD+(0,1)];
```



- **Slices can also be expressed with ranges:**

```
A[2..3, ..] = B[3.., 1..n];
```

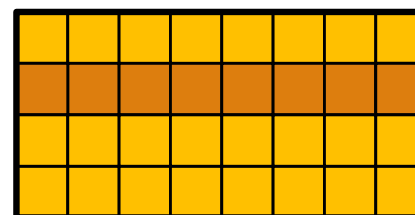




# Rank Change Slicing

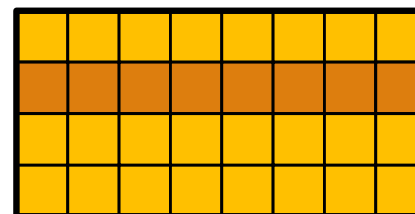
- **Slicing using a 1-element range preserves dimensionality**
  - This is a 2D array expression that's 1 x n:

`...A[2..2, ..]...`



- **Slicing using a scalar results in a rank change:**
  - This is a 1D array expression of  $n$  elements:

`...A[2, ..]...`



# Jacobi Iteration in Chapel

```
config const n = 6,
            epsilon = 1.0e-5;
```

## Compute maximum change

**op reduce**  $\Rightarrow$  collapse aggregate expression to scalar using *op*

**Promotion:** *abs()* and  $-$  are scalar operators; providing array operands results in parallel evaluation equivalent to:

```
forall (a,t) in zip(A,Temp) do abs(a - t)
```

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

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

writeln(A);
```

# Reductions in Chapel

- **Standard reductions supported by default:**

`+, *, min, max, &, |, &&, ||, minloc, maxloc, ...`

- **Reductions can reduce arbitrary iterable expressions:**

```
const total = + reduce Arr,
      factN = * reduce 1..n,
      biggest = max reduce (for i in myIter() do foo(i));
```

- **Advanced users can write their own reductions**

- However, note that the interface is still evolving

# Jacobi Iteration in Chapel

```
config const n = 6,  
            epsilon = 1.0e-5;
```

```
const BigD = {0..n+1, 0..n+1},  
            D = BigD[1..n, 1..n],
```

## Copy data back & Repeat until done

uses slicing and whole array assignment  
standard *do...while* loop construct

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

# Jacobi Iteration in Chapel

```
config const n = 6,  
            epsilon = 1.0e-5;
```

```
const BigD = {0..n+1, 0..n+1},  
            D = BigD[1..n, 1..n],  
            LastRow = D.exterior(1,0);
```

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

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

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

```
writeln(A);
```

Write array to console

# Jacobi Iteration in Chapel

```

config const n = 6,
               epsilon = 1.0e-5;

const BigD = {0..n+1, 0..n+1},
        D = BigD[1..n, 1..n],
        LastRow = D.exterior(1,0);

var A, Temp : [BigD] real;

A[LastRow] = 1.0;

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

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

writeln(A);

use BlockDist;

```

# Jacobi Iteration in Chapel

```
config const n = 6,  
            epsilon = 1.0e-5;
```

```
const BigD = {0..n+1, 0..n+1},  
            D = BigD[1..n, 1..n],  
            LastRow = D.exterior(1,0);
```

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

By default, domains and their arrays are mapped to a single locale.  
Any data parallelism over such domains/ arrays will be executed by the cores on that locale.  
Thus, this is a shared-memory parallel program.

```
Temp[i,j] = (A[i-1,j] + A[i+1,j] + A[i,j-1] + A[i,j+1]) / 4;
```

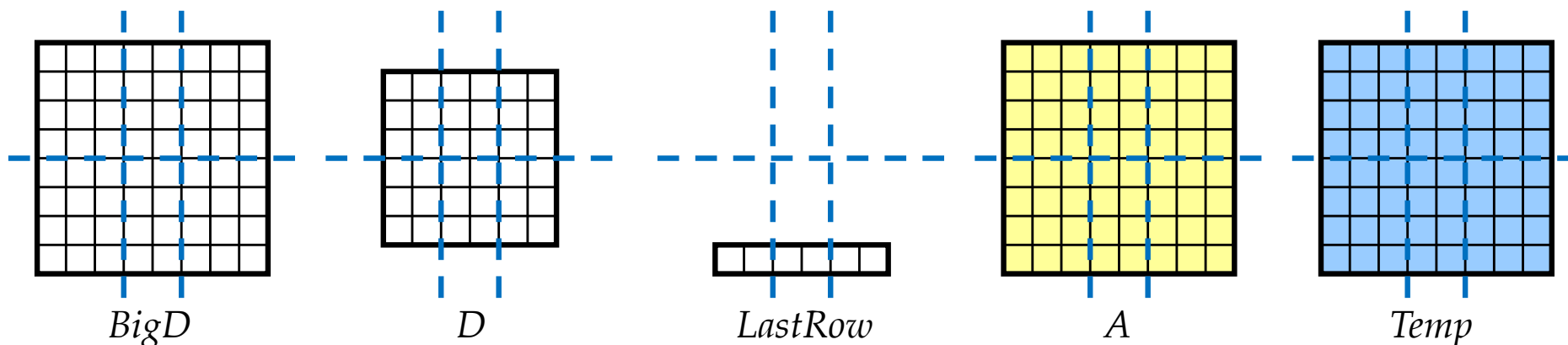
```
const delta = max reduce abs(A[D] - Temp[D]);  
A[D] = Temp[D];  
} while (delta > epsilon);
```

```
writeln(A);
```

# Jacobi Iteration in Chapel (distributed memory)

```
config const n = 6,  
            epsilon = 1.0e-5;  
  
const BigD = {0..n+1, 0..n+1} dmapped Block({1..n, 1..n}),  
            D = BigD[1..n, 1..n],  
            LastRow = D.exterior(1,0);
```

With these simple changes, we specify a mapping from the domains and arrays to locales  
Domain maps describe the mapping of domain indices and array elements to *locales*  
specifies how array data is distributed across locales  
specifies how iterations over domains/arrays are mapped to locales



```
use BlockDist;
```



# Jacobi Iteration in Chapel (distributed memory)

```

config const n = 6,
              epsilon = 1.0e-5;

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

var A, Temp : [BigD] real;

A[LastRow] = 1.0;

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

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

writeln(A);

use BlockDist;

```

# Questions about Data Parallelism?



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