

Chapel

the Cascade High Productivity Language

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Chapel

Chapel: a new parallel language being developed by Cray Inc.

Themes:

- **general parallel programming**
 - data-, task-, and nested parallelism
 - express general levels of software parallelism
 - target general levels of hardware parallelism
- ***global-view* abstractions**
 - stay tuned...
- ***multiresolution* design**
 - program abstractly or close to the machine as needed
- **control of locality**
 - necessary for scalability
- **reduce gap between mainstream & parallel languages**

Chapel's Setting: HPCS

HPCS: High *Productivity* Computing Systems (DARPA *et al.*)

- **Goal:** Raise HEC user productivity by 10× for the year 2010
- **Productivity** = Performance
 - + Programmability
 - + Portability
 - + Robustness
- **Phase II:** Cray, IBM, Sun (July 2003 – June 2006)
 - Evaluated the entire system architecture's impact on productivity...
 - processors, memory, network, I/O, OS, runtime, compilers, tools, ...
 - ...and new languages:
 - Cray:** Chapel
 - IBM:** X10
 - Sun:** Fortress
- **Phase III:** Cray, IBM (July 2006 – 2010)
 - Implement the systems and technologies resulting from phase II
 - (Sun also continues work on Fortress, without HPCS funding)

Chapel and Productivity

Chapel's Productivity Goals:

- vastly improve **programmability** over current languages/models
 - writing parallel codes
 - reading, modifying, porting, tuning, maintaining, evolving them
- support **performance** at least as good as MPI
 - competitive with MPI on generic clusters
 - better than MPI on more capable architectures
- improve **portability** compared to current languages/models
 - as ubiquitous as MPI, but with fewer architectural assumptions
 - more portable than OpenMP, UPC, CAF, ...
- improve **code robustness** via improved semantics and concepts
 - eliminate common error cases altogether
 - better abstractions to help avoid other errors

Outline

- ✓ Chapel Context
- Global-view Programming Models
- Language Overview
- Example Computations
- Status, Future Work, Collaborations

Parallel Programming Model Taxonomy

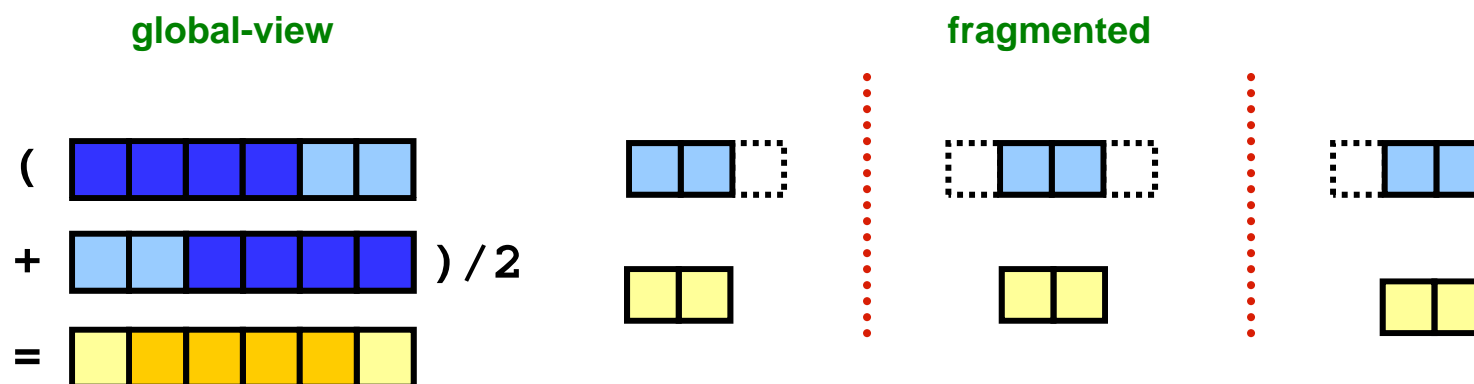
programming model: the mental model a programmer uses when coding using a language, library, or other notation

fragmented models: those in which the programmer writes code from the point-of-view of a single processor/thread

global-view models: those in which the programmer can write code that describes the computation as a whole

Global-view vs. Fragmented

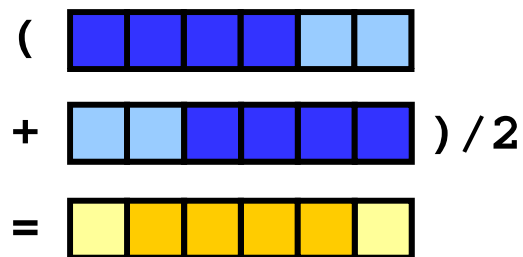
Problem: “Apply 3-pt stencil to vector”



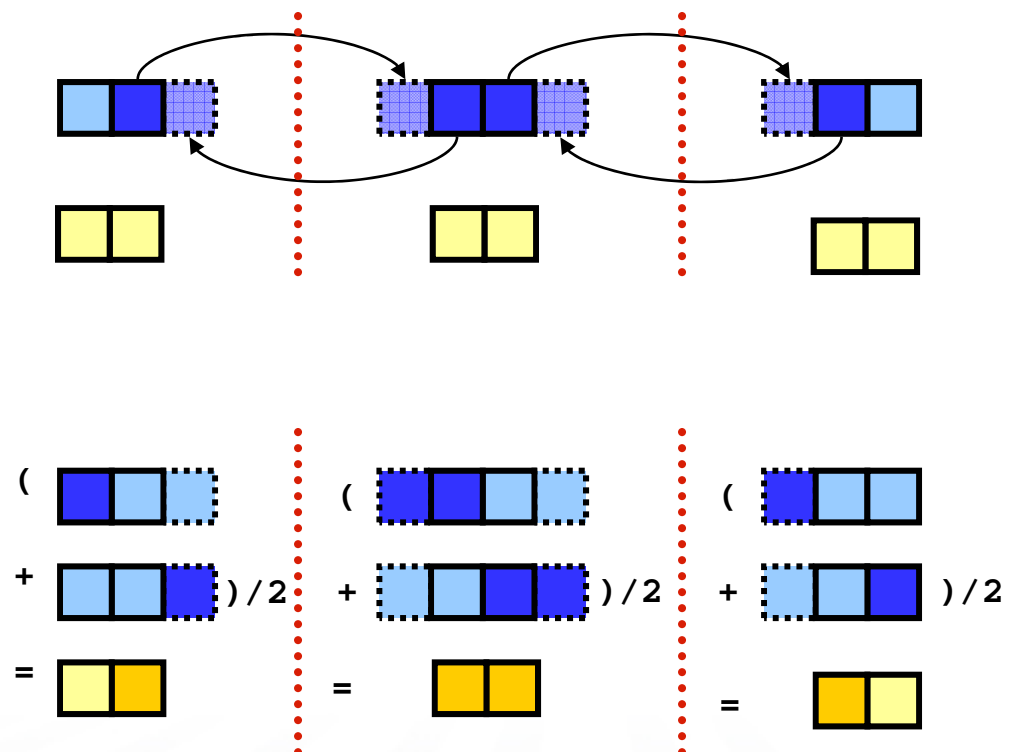
Global-view vs. Fragmented

Problem: “Apply 3-pt stencil to vector”

global-view



fragmented



Parallel Programming Model Taxonomy

programming model: the mental model a programmer uses when coding using a language, library, or other notation

fragmented models: those in which the programmer writes code from the point-of-view of a single processor/thread


SPMD models: Single-Program, Multiple Data -- a common fragmented model in which the user writes one program & runs multiple copies of it, parameterized by a unique ID

global-view models: those in which the programmer can write code that describes the computation as a whole

Global-view vs. SPMD Code

Problem: “Apply 3-pt stencil to vector”

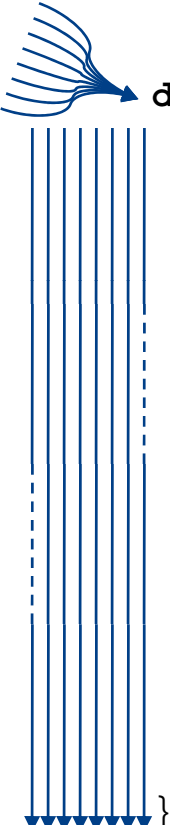
global-view



```
def main() {
  var n: int = 1000;
  var a, b: [1..n] real;

  forall i in 2..n-1 {
    b(i) = (a(i-1) + a(i+1))/2;
  }
}
```

SPMD



```
def main() {
  var n: int = 1000;
  var locN: int = n/numProcs;
  var a, b: [0..locN+1] real;

  if (iHaveRightNeighbor) {
    send(right, a(locN));
    recv(right, a(locN+1));
  }
  if (iHaveLeftNeighbor) {
    send(left, a(1));
    recv(left, a(0));
  }

  forall i in 1..locN {
    b(i) = (a(i-1) + a(i+1))/2;
  }
}
```

Global-view vs. SPMD Code


Problem: “Apply 3-pt stencil to vector”

Assumes *numProcs* divides *n*;
a more general version would
require additional effort

global-view

```
def main() {
  var n: int = 1000;
  var a, b: [1..n] real;

  forall i in 2..n-1 {
    b(i) = (a(i-1) + a(i+1))/2;
  }
}
```



A diagram illustrating the global-view stencil application. It shows a vertical vector of size *n*. A central point is highlighted, and arrows indicate the stencil operation: $b(i) = (a(i-1) + a(i+1))/2$. The operation is applied to all elements from index 2 to *n*-1.

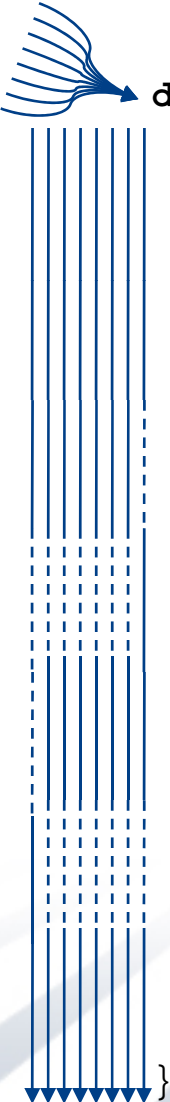
SPMD

```
def main() {
  var n: int = 1000;
  var locN: int = n/numProcs;
  var a, b: [0..locN+1] real;
  var innerLo: int = 1;
  var innerHi: int = locN;

  if (iHaveRightNeighbor) {
    send(right, a(locN));
    recv(right, a(locN+1));
  } else {
    innerHi = locN-1;
  }

  if (iHaveLeftNeighbor) {
    send(left, a(1));
    recv(left, a(0));
  } else {
    innerLo = 2;
  }

  forall i in innerLo..innerHi {
    b(i) = (a(i-1) + a(i+1))/2;
  }
}
```



A diagram illustrating the SPMD stencil application. It shows a vertical vector of size *n* divided into segments. Arrows indicate the distribution of the stencil operation across the segments. The operation is applied to all elements from index 2 to *n*-1, with boundary conditions handled by the *if* statements.

Current HPC Programming Notations

■ communication libraries:

- MPI, MPI-2
- SHMEM, ARMCI, GASNet

(fragmented, typically SPMD)
(SPMD)

■ shared memory models:

- OpenMP

(global-view, trivially)

■ PGAS languages:

- Co-Array Fortran
- UPC
- Titanium

(SPMD)
(SPMD)
(SPMD)

MPI SPMD pseudo-code

Problem: “Apply 3-pt stencil to vector”

SPMD (pseudocode + MPI)

```
var n: int = 1000, locN: int = n/numProcs;
var a, b: [0..locN+1] real;
var innerLo: int = 1, innerHi: int = locN;
var numProcs, myPE: int;
var retval: int;
var status: MPI_Status;

MPI_Comm_size(MPI_COMM_WORLD, &numProcs);
MPI_Comm_rank(MPI_COMM_WORLD, &myPE);
if (myPE < numProcs-1) {
    retval = MPI_Send(&a(locN)), 1, MPI_FLOAT, myPE+1, 0, MPI_COMM_WORLD);
    if (retval != MPI_SUCCESS) { handleError(retval); }
    retval = MPI_Recv(&a(locN+1)), 1, MPI_FLOAT, myPE+1, 1, MPI_COMM_WORLD, &status);
    if (retval != MPI_SUCCESS) { handleErrorWithStatus(retval, status); }
} else
    innerHi = locN-1;
if (myPE > 0) {
    retval = MPI_Send(&a(1)), 1, MPI_FLOAT, myPE-1, 1, MPI_COMM_WORLD);
    if (retval != MPI_SUCCESS) { handleError(retval); }
    retval = MPI_Recv(&a(0)), 1, MPI_FLOAT, myPE-1, 0, MPI_COMM_WORLD, &status);
    if (retval != MPI_SUCCESS) { handleErrorWithStatus(retval, status); }
} else
    innerLo = 2;
forall i in (innerLo..innerHi) {
    b(i) = (a(i-1) + a(i+1))/2;
}
```

Communication becomes
geometrically more complex for
higher-dimensional arrays

Fortran+MPI 3D 27-point stencil (NAS MG *rprj3*)

```

subroutine comm3(u,n1,n2,n3,kk)
  use caf_intrinsics

  implicit none
  include 'cafnpb.h'
  include 'globals.h'

  integer n1, n2, n3, kk
  double precision u(n1,n2,n3)
  integer axis

  if( .not. dead(kk) )then
    do axis = 1, 3
      if( nprocx .ne. 1 ) then
        call sync_all()
        call give3( axis, +1, u, n1, n2, n3, kk )
        call give3( axis, -1, u, n1, n2, n3, kk )
        call sync_all()
        call take3( axis, -1, u, n1, n2, n3 )
        call take3( axis, +1, u, n1, n2, n3 )
      else
        call comm3p( axis, u, n1, n2, n3, kk )
      endif
    enddo
  else
    do axis = 1, 3
      call sync_all()
      call sync_all()
    enddo
    call zero3(u,n1,n2,n3)
  endif
  return
end

subroutine give3( axis, dir, u, n1, n2, n3, k )
  use caf_intrinsics

  implicit none
  include 'cafnpb.h'
  include 'globals.h'

  integer axis, dir, n1, n2, n3, k, ierr
  double precision u( n1, n2, n3 )
  integer i3, i2, i1, buff_len, buff_id

  buff_id = 2 + dir
  buff_len = 0

  if( axis .eq. 1 )then
    if( dir .eq. -1 )then
      do i3=2,n3-1
        do i2=2,n2-1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( 2, i2,i3)
        enddo
      enddo
    > buff(1:buff_len, buff_id+1)[nbr(axis,dir,k)] =
      buff(1:buff_len, buff_id)

    else if( dir .eq. +1 ) then
      do i3=2,n3-1
        do i2=2,n2-1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( n1-1, i2,i3)
        enddo
      enddo
    > buff(1:buff_len, buff_id+1)[nbr(axis,dir,k)] =
      buff(1:buff_len, buff_id)

    endif
  endif

  if( axis .eq. 2 )then
    if( dir .eq. -1 )then
      do i3=2,n3-1
        do i1=1,n1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( i1, 2,i3)
        enddo
      enddo
    > buff(1:buff_len, buff_id+1)[nbr(axis,dir,k)] =
      buff(1:buff_len, buff_id)

    else if( dir .eq. +1 ) then
      do i3=2,n3-1
        do i1=1,n1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( i1,i2,2)
        enddo
      enddo
    > buff(1:buff_len, buff_id+1)[nbr(axis,dir,k)] =
      buff(1:buff_len, buff_id)

    else if( dir .eq. -1 ) then
      do i3=2,n3-1
        do i1=1,n1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( i1,i2,n3)
        enddo
      enddo
    > buff(1:buff_len, buff_id+1)[nbr(axis,dir,k)] =
      buff(1:buff_len, buff_id)

    else if( dir .eq. +1 ) then
      do i3=2,n3-1
        do i1=1,n1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( i1,i2,n3)
        enddo
      enddo
    > buff(1:buff_len, buff_id+1)[nbr(axis,dir,k)] =
      buff(1:buff_len, buff_id)

    endif
  endif

  return
end

subroutine take3( axis, dir, u, n1, n2, n3 )
  implicit none
  include 'cafnpb.h'
  include 'globals.h'

  integer axis, dir, n1, n2, n3
  double precision u( n1, n2, n3 )
  integer i3, i2, i1

  buff_id = 3 + dir
  indx = 0

  if( axis .eq. 1 )then
    if( dir .eq. -1 )then
      do i3=2,n3-1
        do i2=2,n2-1
          indx = indx + 1
          u(n1,i2,i3) = buff(indx, buff_id )
        enddo
      enddo
    else if( dir .eq. +1 ) then
      do i3=2,n3-1
        do i2=2,n2-1
          indx = indx + 1
          u(1,i2,i3) = buff(indx, buff_id )
        enddo
      enddo
    endif
  endif

  if( axis .eq. 2 )then
    if( dir .eq. -1 )then
      do i3=2,n3-1
        do i1=1,n1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( n1-1,
            1,i3)
        enddo
      enddo
    > buff(1:buff_len, buff_id) = u( n1-1,
      1,i3)

    else if( dir .eq. +1 ) then
      do i3=2,n3-1
        do i1=1,n1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( i1,n2-1,
            1,i3)
        enddo
      enddo
    > buff(1:buff_len, buff_id) = u( i1,n2-1,
      1,i3)

    endif
  endif

  if( axis .eq. 3 )then
    if( dir .eq. -1 )then
      do i2=1,n2
        do i1=1,n1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( i1,i2,n3)
        enddo
      enddo
    > buff(1:buff_len, buff_id) = u( i1,i2,n3)

    else if( dir .eq. +1 ) then
      do i2=1,n2
        do i1=1,n1
          buff_len = buff_len + 1
          buff(buff_len, buff_id) = u( i1,i2,1)
        enddo
      enddo
    > buff(1:buff_len, buff_id) = u( i1,i2,1)

    endif
  endif

  return
end

subroutine comm3p( axis, u, n1, n2, n3, kk )
  use caf_intrinsics

  implicit none
  include 'cafnpb.h'
  include 'globals.h'

  integer axis, dir, n1, n2, n3
  double precision u( n1, n2, n3 )
  integer i3, i2, i1, buff_len, buff_id
  integer i, kk, indx

  dir = -1
  buff_id = 3 + dir
  buff_len = nm2

  do i=1,nm2
    buff(1, buff_id) = 0.0D0
  enddo

  dir = +1
  buff_id = 3 + dir
  buff_len = nm2

  do i=1,nm2
    buff(1, buff_id) = 0.0D0
  enddo

  dir = +1
  buff_id = 2 + dir
  buff_len = 0

  if( axis .eq. 1 )then
    do i3=2,n3-1
      do i2=2,n2-1
        buff_len = buff_len + 1
        buff(buff_len, buff_id) = u( n1-1,
          i2,i3)
      enddo
    enddo
  endif

  if( axis .eq. 2 )then
    do i3=2,n3-1
      do i1=1,n1
        buff_len = buff_len + 1
        buff(buff_len, buff_id) = u( i1,i2,i3)
      enddo
    enddo
  endif

  if( axis .eq. 3 )then
    do i2=1,n2
      do i1=1,n1
        buff_len = buff_len + 1
        buff(buff_len, buff_id) = u( i1,i2,i3)
      enddo
    enddo
  endif

  return
end

subroutine rprj3(r,mlk,m2k,m3k,s,m1j,m2j,m3j,k)
  implicit none
  include 'cafnpb.h'
  include 'globals.h'

  integer mlk, m2k, m3k, m1j, m2j, m3j, k

  double precision r( mlk,m2k,m3k ), s( m1j,m2j,m3j )
  integer j3, j2, j1, i3, i2, i1, d1, d2, d3, j
  double precision x1(m), y1(m), x2,y2

  if( mlk.eq.3 )then
    d1 = 2
  else
    d2 = 1
  endif

  if( m2k.eq.3 )then
    d2 = 2
  else
    d2 = 1
  endif

  if( m3k.eq.3 )then
    d3 = 2
  else
    d3 = 1
  endif

  do j3=2,m3j-1
    i3 = 2*j3-d3
    do j2=2,m2j-1
      i2 = 2*j2-d2
      do j1=2,m1j
        i1 = 2*j1-d1
        x1(i1-1) = r( i1-1,i2-1,i3 ) + r( i1-1,i2+1,i3 )
        > + r( i1-1,i2, i3-1 ) + r( i1-1,i2, i3+1 )
        y1(i1-1) = r( i1-1,i2-1,i3-1 ) + r( i1-1,i2-1,i3+1 )
        > + r( i1-1,i2+1,i3-1 ) + r( i1-1,i2+1,i3+1 )
      enddo
    enddo
    do j1=2,m1j-1
      i1 = 2*j1-d1
      y2 = r( i1, i2-1,i3-1 ) + r( i1, i2-1,i3+1 )
      > + r( i1, i2+1,i3-1 ) + r( i1, i2+1,i3+1 )
      x2 = r( i1, i2-1,i3 ) + r( i1, i2+1,i3 )
      > + r( i1, i2, i3-1 ) + r( i1, i2, i3+1 )
      s( j1,j2,j3 ) =
      > 0.5D0 * r( i1,i2,i3 )
      > + 0.25D0 * ( r( i1-1,i2,i3 ) + r( i1+1,i2,i3 ) + x2 )
      > + 0.125D0 * ( x1(i1-1) + x1(i1+1) + y2 )
      > + 0.0625D0 * ( y1(i1-1) + y1(i1+1) )
    enddo
  enddo
  call comm3(s,m1j,m2j,m3j,j)
  return
end

```

Summarizing Fragmented/SPMD Models

■ Advantages:

- fairly straightforward model of execution
- relatively easy to comprehend, learn, reason about
- relatively easy to implement
- reasonable performance on commodity architectures
- portable/ubiquitous
- lots of important scientific work has been accomplished using them

■ Disadvantages:

- blunt means of expressing parallelism: cooperating executables
- fails to abstract away architecture / implementing mechanisms
- obfuscates algorithms with many low-level details
 - error-prone
 - brittle code: difficult to read, maintain, modify, *experiment*
 - “MPI: the assembly language of parallel computing”

Multiresolution Language Design

Conventional Wisdom: By providing higher-level concepts in a language, programmers' hands are tied, preventing them from optimizing for performance by hand

My Belief: With appropriate design, this need not be the case

- provide high-level features and automation for convenience
 - knowledge of such features can aid in compiler optimization
 - provide capabilities to drop down to lower, more manual levels
 - use appropriate separation of concerns to keep this clean
 - support the 90/10 rule
 - in the limit...
 - ...support interoperability with lower-level languages
 - ...support MPI interface within Chapel (?)
 - ...support ability to inline C/Fortran “assembly” (?)
- (I believe that such capabilities will typically not be needed)

Outline

- ✓ Chapel Context
- ✓ Global-view Programming Models
 - ZPL Detour?
- Language Overview
 - Base Language
 - Parallel Features
 - task parallel
 - data parallel
 - Locality Features
- Example Computations
- Status, Future Work, Collaborations

Base Language: Themes

- Block-structured, imperative programming
- Intentionally not an extension to an existing language
- Instead, select attractive features from others:
 - ZPL, HPF:** data parallelism, index sets, distributed arrays
(see also APL, NESL, Fortran90)
 - Cray MTA C/Fortran:** task parallelism, lightweight synchronization
 - CLU:** iterators (see also Ruby, Python, C#)
 - ML:** latent types (see also Scala, Matlab, Perl, Python, C#)
 - Java, C#:** OOP, type safety
 - C++:** generic programming/templates (without adopting its syntax)
 - C, Modula, Ada:** syntax

Base Language: Overview

- Syntax
 - adopt C/C++/Java/Perl syntax whenever possible/useful
 - main departures: declarations/casts, for loops, generics
- Language Elements
 - standard scalar types, expressions, statements
 - value- and reference-based OOP (optional)
 - no pointers, restricted opportunities for aliasing
 - argument intents similar to Fortran/Ada
- My favorite base language features
 - rich compile-time language
 - latent types / simple static type inference
 - configuration variables
 - tuples
 - iterators...

Base Language: Standard Stuff

- Lexical structure and syntax based largely on C/C++

```
{ a = b + c;  foo(); }
```

// no surprises here
- Reasonably standard in terms of:
 - scalar types
 - constants, variables
 - operators, expressions, statements, functions
- Support for object-oriented programming
 - value- and reference-based classes
 - no strong requirement to use OOP
- Modules for namespace management
- Generic functions and classes

Base Language: Departures

- **Syntax:** declaration syntax differs from C/C++

```
var <varName> [ : <definition> ] [ = <init> ] ;
```

```
def <fnName> ( <argList> ) [ : <returnType> ] { ... }
```

- **Types**

- support for complex, imaginary, string types
- sizes more explicit than in C/C++ (e.g., **int**(32), **complex**(128))
- richer array support than C/C++, Java, even Fortran
- no pointers (apart from class references)

- **Operators**

- casts via ':' (e.g., 3.14: **int**(32))
- exponentiation via '**' (e.g., 2**n)

- **Statements:** for loop differs from C/C++

```
for <indices> in <iterationSpace> { ... }
```

```
e.g., for i in 1..n { ... }
```

- **Functions:** argument-passing semantics

Base Language: My Favorite Departures

■ Rich compile-time language

- parameter values (compile-time constants)
- folded conditionals, unrolled for loops, expanded tuples
- type and parameter functions – evaluated at compile-time

■ Latent types:

- ability to omit type specifications for convenience or reuse
- type specifications can be omitted from...
 - variables (inferred from initializers)
 - class members (inferred from constructors)
 - function arguments (inferred from callsite)
 - function return types (inferred from return statements)

■ Configuration variables (and parameters)

```
config const n = 100; // override with --n=1000000
```

■ Tuples

■ Iterators...

Base Language: Motivation for Iterators

Given a program with a bunch of similar loops...

```
for (i=0; i<m; i++) {
  for (j=0; j<n; j++) {
    ...A(i,j)...
  }
}
```

...

```
for (i=0; i<m; i++) {
  for (j=0; j<n; j++) {
    ...A(i,j)...
  }
}
```

...

Consider the effort to convert them from RMO to CMO...

```
for (j=0; j<n; j++) {
  for (i=0; i<m; i++) {
    ...A(i,j)...
  }
}
```

...

```
for (j=0; j<n; j++) {
  for (i=0; i<m; i++) {
    ...A(i,j)...
  }
}
```

...

Or to tile the loops...

```
for (jj=0; jj<n; jj+=blocksize) {
  for (ii=0; ii<m; ii+=blocksize) {
    for (j=jj; j<min(m,jj+blocksize-1) {
      for (i=ii; i<min(n,ii+blocksize-1) {
        ...A(i,j)...
      }
    }
  }
}
```

...

```
for (jj=0; jj<n; jj+=blocksize) {
  for (ii=0; ii<m; ii+=blocksize) {
    for (j=jj; j<min(m,jj+blocksize-1) {
      for (i=ii; i<min(n,ii+blocksize-1) {
        ...A(i,j)...
      }
    }
  }
}
```

...

Base Language: Motivation for Iterators

Given a program with a bunch of similar loops...

```
for (i=0; i<m; i++) {
  for (j=0; j<n; j++) {
    ...A(i,j)...
  }
}
```

Consider the effort to convert them from RMO to CMO...

```
for (j=0; j<n; j++) {
  for (i=0; i<m; i++) {
    ...A(i,j)...
  }
}
```

Or to tile the loops...

```
for (jj=0; jj<n; jj+=blocksize) {
  for (ii=0; ii<m; ii+=blocksize) {
    for (j=jj; j<min(m,jj+blocksize-1) {
      for (i=ii; i<min(n,ii+blocksize-1) {
        ...A(i,j)...
      }
    }
  }
}
```

Or to change the iteration order over the tiles...

... Or to make them into fragmented loops for an MPI program...

for Or to change the distribution of the work/arrays in that MPI program...

```
for (i=0; i<n; i++) {
  for (j=0; j<m; j++) {
    ...A(i,j)...
  }
}
```

Or to label them as parallel for OpenMP or a vectorizing compiler...

Or to do *anything* that we do with loops all the time as a community...

We wouldn't program straight-line code this way, so why are we so tolerant of our lack of loop abstractions?

Base Language: Iterators

- like functions, but *yield* a number of elements one-by-one:

```

iterator RMO() {
    for i in 1..m do
        for j in 1..n do
            yield (i,j);
        }
    }

iterator tiled(blocksize) {
    for ii in 1..m by blocksize do
        for jj in 1..n by blocksize do
            for i in ii..min(n, ii+blocksize-1) do
                for j in jj..min(m, jj+blocksize-1) {
                    yield (i,j);
                }
            }
        }
    }

```

- iterators are used to drive loops:

```

for ij in RMO() {
    ...A(ij)...
}

for ij in tiled(blocksize) {
    ...A(ij)...
}

```

- as with functions...
 - ...one iterator can be redefined to change the behavior of many loops
 - ...a single invocation can be altered, or its arguments can be changed
- not necessarily any more expensive than in-line loops

Task Parallelism: Task Creation

begin: creates a task for future evaluation

```
begin DoThisTask();  
WhileContinuing();  
TheOriginalThread();
```

cobegin: creates a task per component statement:

```
computePivot(lo, hi, data);  
cobegin {  
    Quicksort(lo, pivot, data);  
    Quicksort(pivot, hi, data);  
} // implicit join here
```

```
cobegin {  
    computeTaskA(...);  
    computeTaskB(...);  
    computeTaskC(...);  
} // implicit join
```

coforall: creates a task per loop iteration

```
coforall e in Edges {  
    exploreEdge(e);  
} // implicit join here
```

Task Parallelism: Task Coordination

sync variables: store full/empty state along with value

```
var result$: sync real;    // result is initially empty
cobegin {
    ... = result$;         // block until full, leave empty
    result$ = ...;         // block until empty, leave full
}
result$.readFF();          // read when full, leave full;
                          // other variations also supported
```

single-assignment variables: writable once only

```
var result$: single real = begin f(); // result initially empty
...                                  // do some other things
total += result$;                   // block until result has been filled
```

atomic sections: support transactions against memory

```
atomic {
    newnode.next = insertpt;
    newnode.prev = insertpt.prev;
    insertpt.prev.next = newnode;
    insertpt.prev = newnode;
}
```

Producer/Consumer example

```
var buff$: [0..bufferSize-1] sync int;
```

```
cobegin {  
    producer();  
    consumer();  
}
```

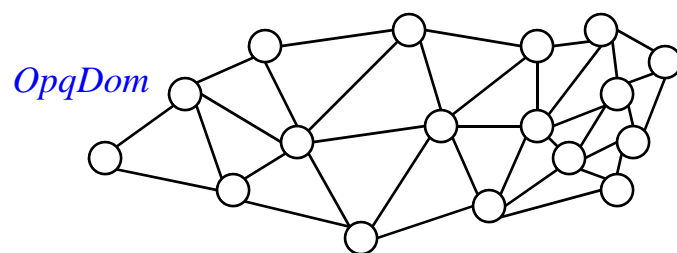
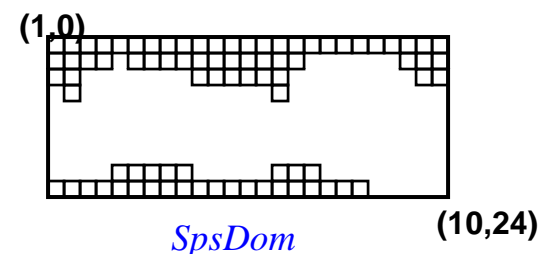
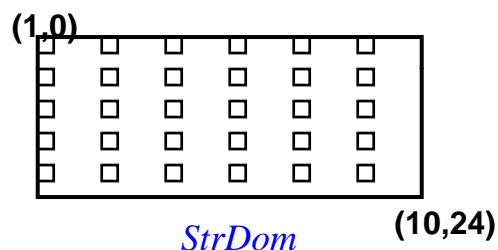
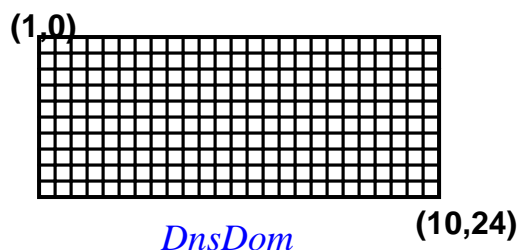
```
def producer() {  
    var i = 0;  
    for ... {  
        i = (i+1) % bufferSize;  
        buff$(i) = ...;  
    }  
}
```

```
def consumer() {  
    var i = 0;  
    while {  
        i = (i+1) % bufferSize;  
        ...buff$(i)...;  
    }  
}
```

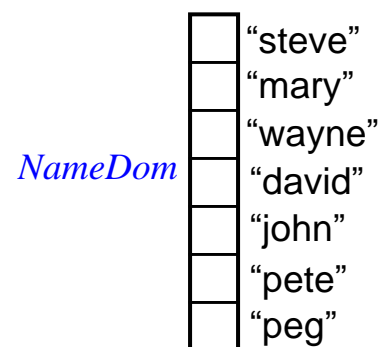
Data Parallelism: Domains

domains: first-class index sets, whose indices can be...

...integer tuples



...anonymous

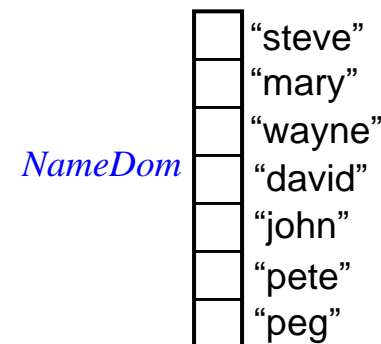
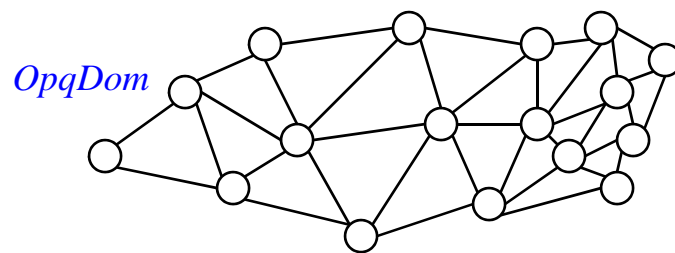
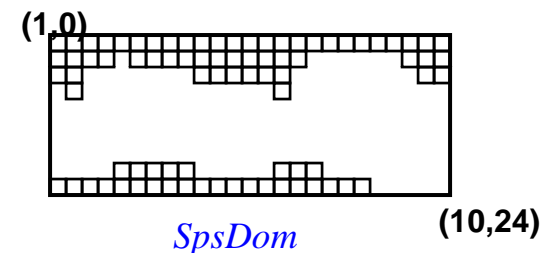
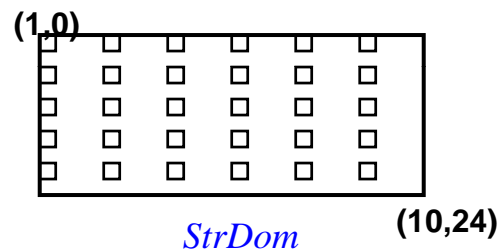
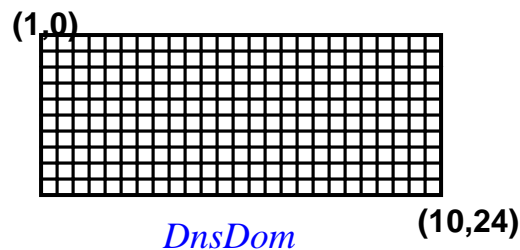


...arbitrary values

Data Parallelism: Domain Declarations

domains: first-class index sets, whose indices can be...

```
var DnsDom: domain(2) = [1..10, 0..24],
    StrDom: subdomain(DnsDom) = DnsDom by (2,4),
    SpsDom: subdomain(DnsDom) = genIndices();
```



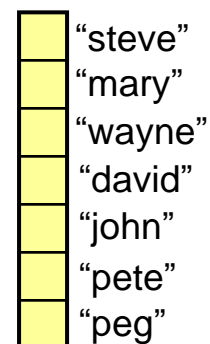
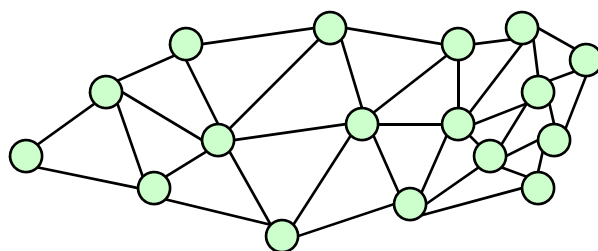
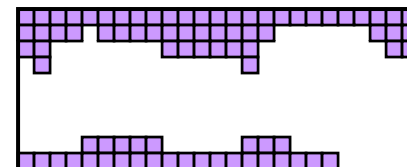
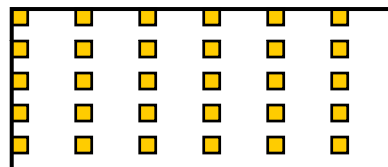
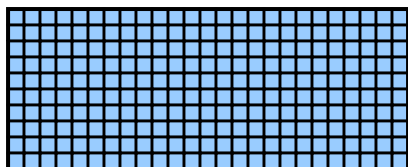
```
var OpqDom: domain(opaque),
    NameDom: domain(string) = readNames();
```

Data Parallelism: Domains and Arrays

Domains are used to declare arrays...

```
var DnsArr: [DnsDom] complex,
    SpsArr: [SpsDom] real;
```

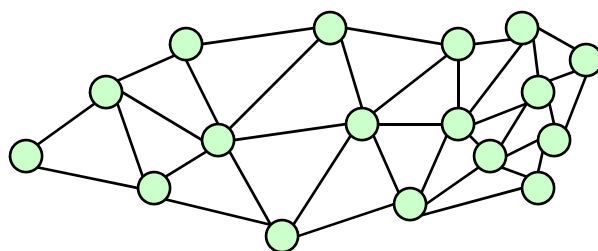
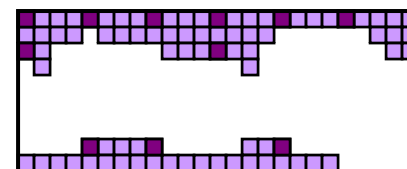
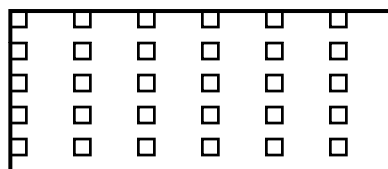
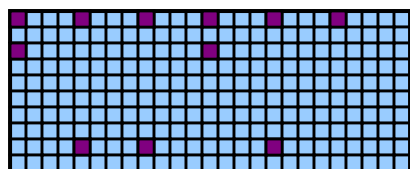
...



Data Parallelism: Domain Iteration

...to iterate over index spaces...

```
forall (i,j) in StrDom {
  DnsArr(i,j) += SpsArr(i,j);
}
```

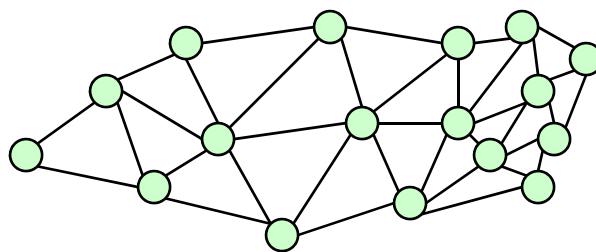
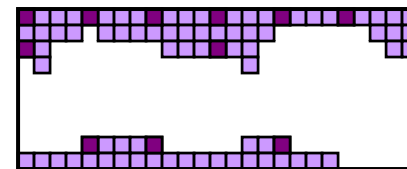
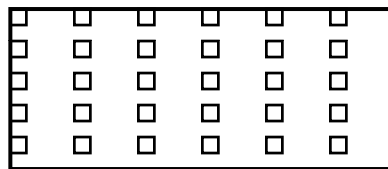
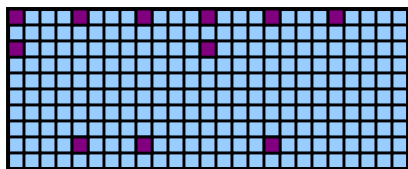


	"steve"
	"mary"
	"wayne"
	"david"
	"john"
	"pete"
	"peg"

Data Parallelism: Array Slicing

...to slice arrays...

```
DnsArr[StrDom] += SpsArr[StrDom];
```



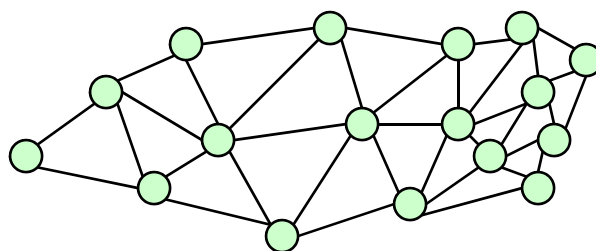
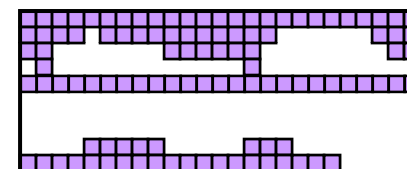
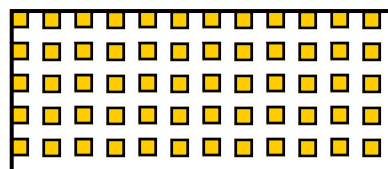
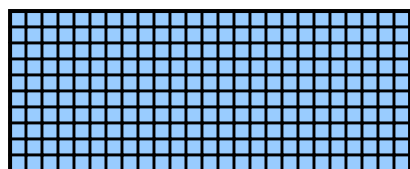
	"steve"
	"mary"
	"wayne"
	"david"
	"john"
	"pete"
	"peg"

Data Parallelism: Array Reallocation

...and to reallocate arrays

```
StrDom = DnsDom by (2,2);
```

```
SpsDom += genEquator();
```



“steve”
“mary”
“wayne”
“david”
“john”
“pete”
“peg”

Locality: Locales

- *locale*: architectural unit of storage and processing
- user specifies # locales on executable command-line

```
prompt> myChapelProg -nl=8
```
- Chapel programs have a built-in domain/array of locales:

```
const LocaleSpace: domain(1) = [0..numLocales-1];
const Locales: [LocaleSpace] locale;
```

- Users can use this to create their own locale arrays:

```
var CompGrid: [1..GridRows, 1..GridCols] locale = ...;
```

0	1	2	3
4	5	6	7

CompGrid

```
var TaskALocs = Locales[1..numTaskALocs];
```

```
var TaskBLocs = Locales[1..numTaskBLocs];
```

0	1
---	---

TaskALocs

2	3	4	5	6	7
---	---	---	---	---	---

TaskBLocs

Locality: Task Placement

on clauses: indicate where tasks should execute

Either in a data-driven manner...

```
computePivot(lo, hi, data);
cobegin {
  on A(lo)      do Quicksort(lo, pivot, data);
  on A(pivot) do Quicksort(pivot, hi, data);
}
```

...or by naming locales explicitly

```
cobegin {
  on Locales(0) do producer();
  on Locales(1) do consumer();
}
```

0	producer()
1	consumer()

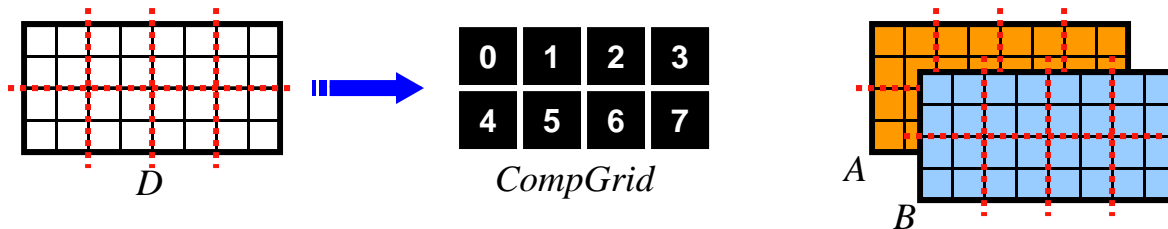
```
cobegin {
  on TaskALocs do computeTaskA(...);
  on TaskBLocs do computeTaskB(...);
  on Locales(0) do computeTaskC(...);
}
```

0	1	computeTaskA()				
2	3	4	5	6	7	computeTaskB()
0	computeTaskC()					

Locality: Domain Distribution

Domains may be distributed across locales

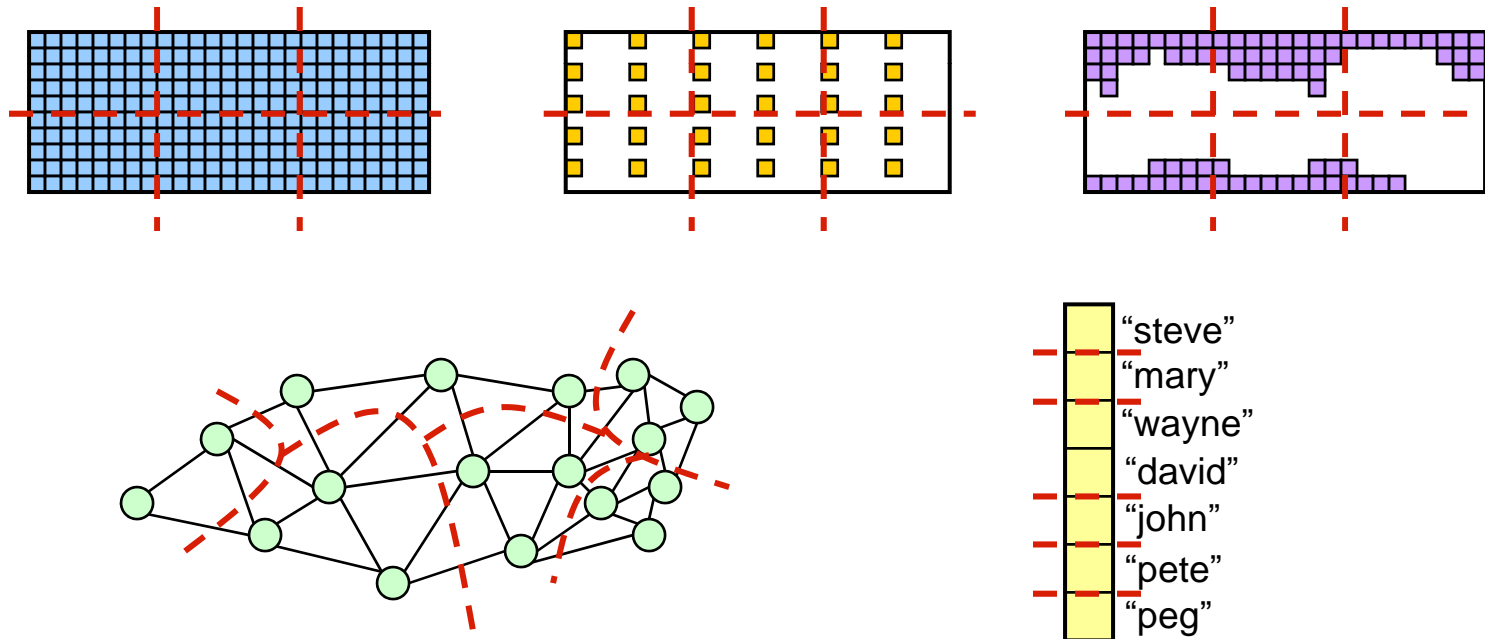
```
var D: domain(2) distributed Block on CompGrid = ...;
```



Locality: Domain Distributions

Distributions specify...

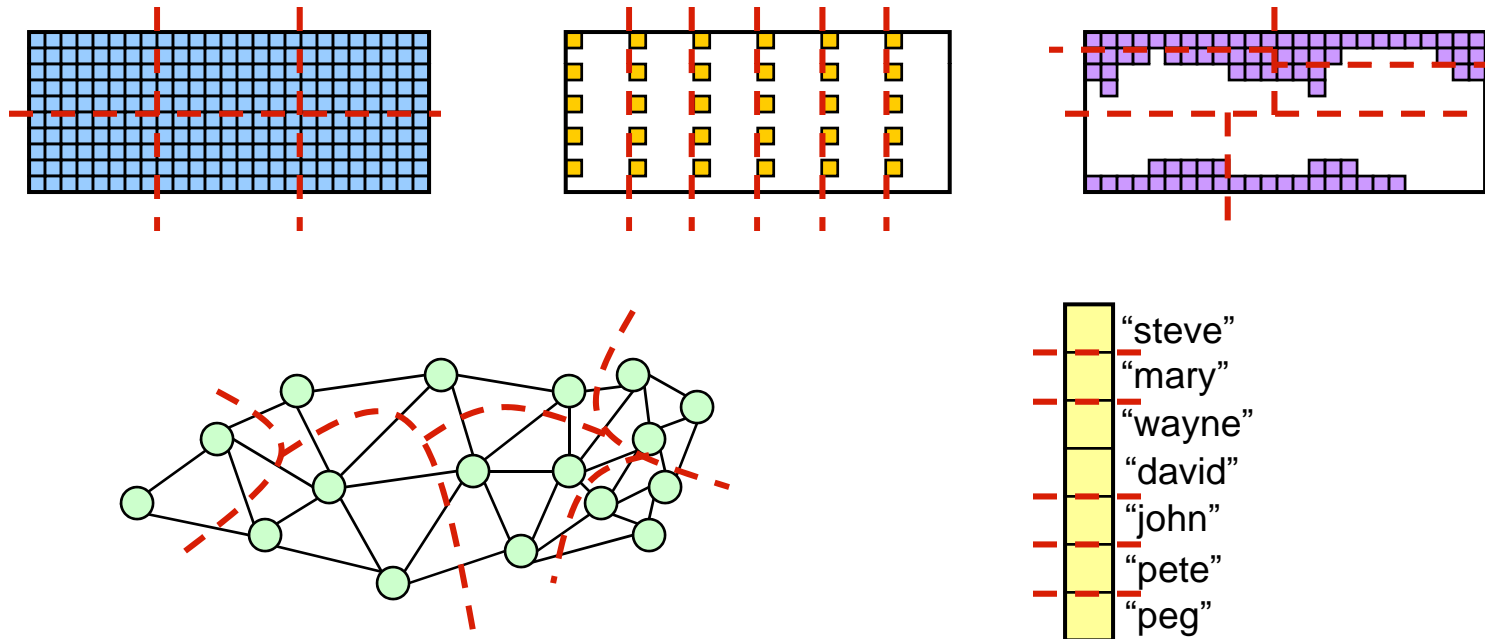
- ...a mapping of indices to locales
- ...per-locale storage for domain indices and array elements
- ...a default work assignment for operations on domains/arrays



Locality: Domain Distributions

Distributions specify...

- ...a mapping of indices to locales
- ...per-locale storage for domain indices and array elements
- ...a default work assignment for operations on domains/arrays



Locality: Distributions Overview

Distributions: “recipes for distributed arrays”

- Intuitively, distributions implement the lowering...
 - from:** the user’s global view of distributed data aggregates
 - to:** the fragmented implementation for distributed memory machines
- Users can implement custom distributions
- Author implements an interface which supports:
 - allocation/reallocation of domain indices and array elements
 - mapping functions (e.g., index-to-locale, index-to-value)
 - iterators: parallel/serial; global/local
 - optionally, communication idioms
- Chapel provides a standard library of distributions...
 - ...written using the same mechanism as user-defined distributions
 - ...tuned for different platforms to maximize performance

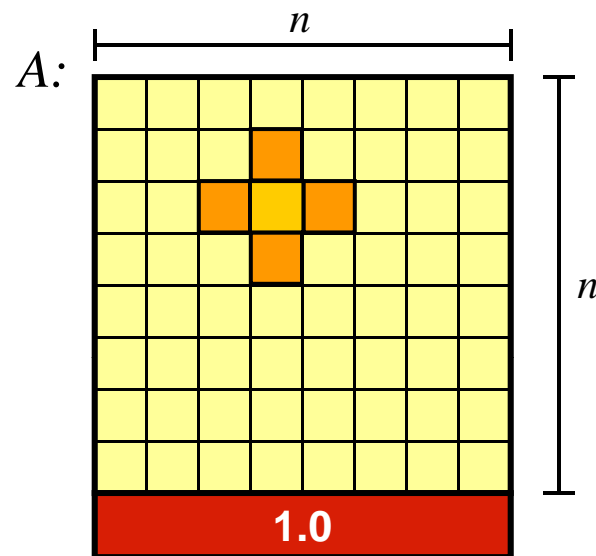
Other Features

- *zippered* and *tensor* flavors of iteration and promotion
- *subdomains* and *index types* to help reason about indices
- *reductions* and *scans* (with user-defined operators)

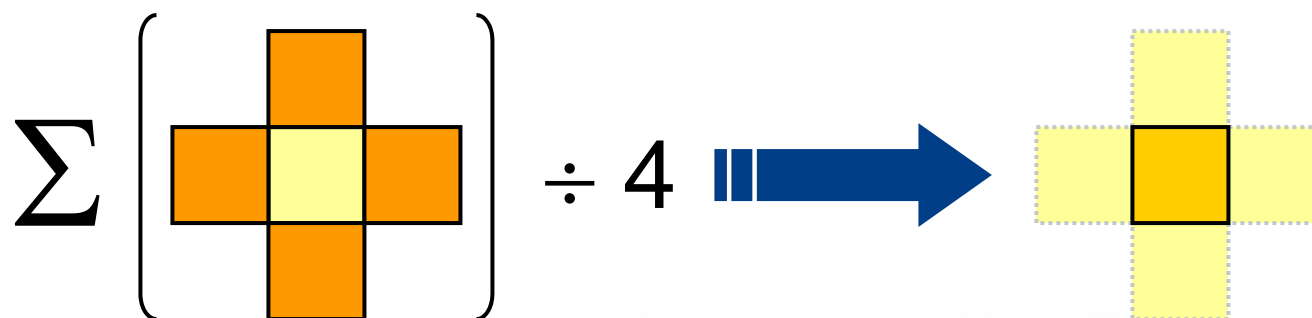
Outline

- ✓ Chapel Context
- ✓ Global-view Programming Models
- ✓ Language Overview
- Example Computations
- Status, Future Work, Collaborations

Example 1: Jacobi Iteration



repeat until max
change $< \epsilon$



Jacobi Iteration in Chapel

```
config const n = 6,  
            epsilon = 1.0e-5;  
  
const BigD: domain(2) = [0..n+1, 0..n+1],  
      D: subdomain(BigD) = [1..n, 1..n],  
      LastRow: subdomain(BigD) = D.exterior(1,0);  
  
var A, Temp : [BigD] real;  
  
A[LastRow] = 1.0;  
  
do {  
  [(i,j) in D] Temp(i,j) = (A(i-1,j) + A(i+1,j)  
                           + A(i,j-1) + A(i,j+1)) / 4.0;  
  
  var 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: domain(2) = [0..n+1, 0..n+1],
      D: subdomain(BigD) = [1..n, 1..n],
      LastRow: subdomain(BigD) = D.exterior(1,0);
```

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

```
A[LastRow]
```

```
do {
  [1..n] i
```

```
var
  A(D)
} while
```

```
writeln(A);
```

Declare program parameters

const ⇒ can't change values after initialization

config ⇒ can be set on executable command-line

prompt> jacobi --n=10000 --epsilon=0.0001

note that no types are given; inferred from initializer

n ⇒ **integer** (current default, 32 bits)

epsilon ⇒ **floating-point** (current default, 64 bits)

Jacobi Iteration in Chapel

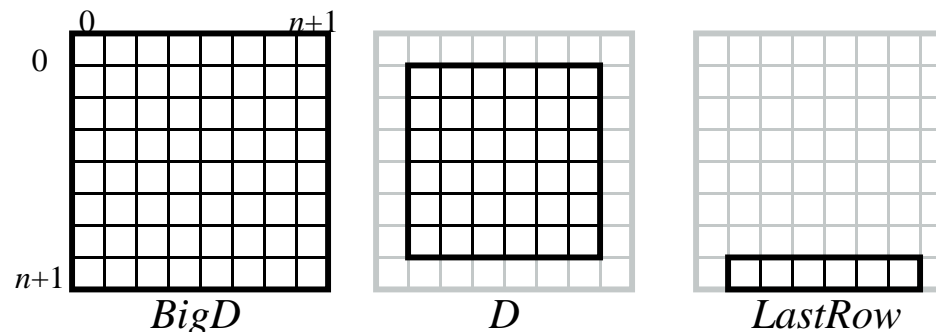
```
config const n = 6,
            epsilon = 1.0e-5;

const BigD: domain(2) = [0..n+1, 0..n+1],
      D: subdomain(BigD) = [1..n, 1..n],
      LastRow: subdomain(BigD) = D.exterior(1,0);
```

Declare domains (first class index sets)

domain(2) \Rightarrow 2D arithmetic domain, indices are integer 2-tuples

subdomain(*P*) \Rightarrow a domain of the same type as *P* whose indices are guaranteed to be a subset of *P*'s



exterior \Rightarrow one of several built-in domain generators

4.0;

Jacobi Iteration in Chapel

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

const BigD: domain(2) = [0..n+1, 0..n+1],
      D: subdomain(BigD) = [1..n, 1..n],
      LastRow: subdomain(BigD) = D.exterior(1,0);

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

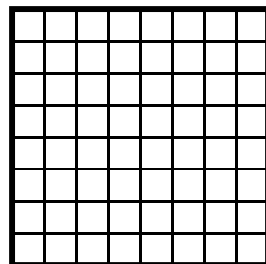
Declare arrays

var \Rightarrow can be modified throughout its lifetime

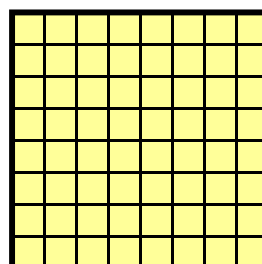
: $T \Rightarrow$ declares variable to be of type T

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

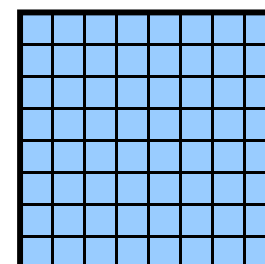
(**no initializer**) \Rightarrow values initialized to default value (0.0 for reals)



BigD



A



Temp

4.0;

Jacobi Iteration in Chapel

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

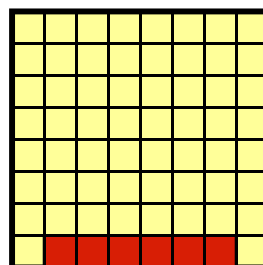
const BigD: domain(2) = [0..n+1, 0..n+1],
      D: subdomain(BigD) = [1..n, 1..n],
      LastRow: subdomain(BigD) = D.exterior(1,0);

var A, Temp : [BigD] real;

A[LastRow] = 1.0;
```

Set Explicit Boundary Condition

indexing by domain \Rightarrow slicing mechanism
array expressions \Rightarrow parallel evaluation



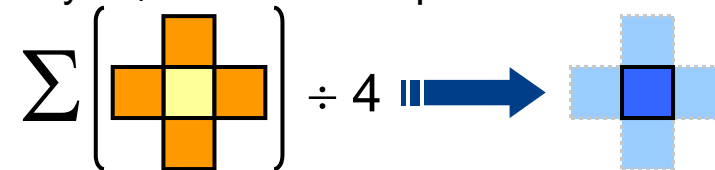
A

Jacobi Iteration in Chapel

Compute 5-point stencil

$[(i,j) \text{ in } D] \Rightarrow$ parallel forall expression over D 's indices, binding them to new variables i and j

Note: since $(i,j) \in D$ and $D \subseteq \text{BigD}$ and $\text{Temp}: [\text{BigD}]$
 \Rightarrow no bounds check required for $\text{Temp}(i,j)$
 with compiler analysis, same can be proven for A 's accesses



```
[(i,j) in D] Temp(i,j) = (A(i-1,j) + A(i+1,j)
                        + A(i,j-1) + A(i,j+1)) / 4.0;
```

```
var 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: domain(2) = [0..n+1, 0..n+1],
```

Compute maximum change

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

Promotion: *abs()* and $-$ are scalar operators, automatically promoted to work with array operands

```
do {
  [(i,j) in D] Temp(i,j) = (A(i-1,j) + A(i+1,j)
                           + A(i,j-1) + A(i,j+1)) / 4.0;

  var 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: domain(2) = [0..n+1, 0..n+1],
      D: subdomain(BigD) = [1..n, 1..n],
      LastRow: subdomain(BigD) = D.exterior(1,0);
```

```
var
```

Copy data back & Repeat until done

A[LastRow] uses slicing and whole array assignment
standard *do...while* loop construct

```
do {
    [(i,j) in D] Temp(i,j) = (A(i-1,j) + A(i+1,j)
                              + A(i,j-1) + A(i,j+1)) / 4.0;

    var 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: domain(2) = [0..n+1, 0..n+1],
      D: subdomain(BigD) = [1..n, 1..n],
      LastRow: subdomain(BigD) = D.exterior(1,0);

var A, Temp : [BigD] real;

A[LastRow] = 1.0;

do {
  [ (i,j) in D ] {
    A(i,j) = (A(i,j-1) + A(i,j+1) + A(i-1,j) + A(i+1,j)) / 4.0;

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

writeln(A);

```

Write array to console

If written to a file, parallel I/O would be used

Jacobi Iteration in Chapel

```

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

const BigD: domain(2) = [0..n+1, 0..n+1] distributed (Block),
      D: subdomain(BigD) = [1..n, 1..n],
      LastRow: subdomain(BigD) = D.exterior(1,0);

var A, Temp : [BigD] real;

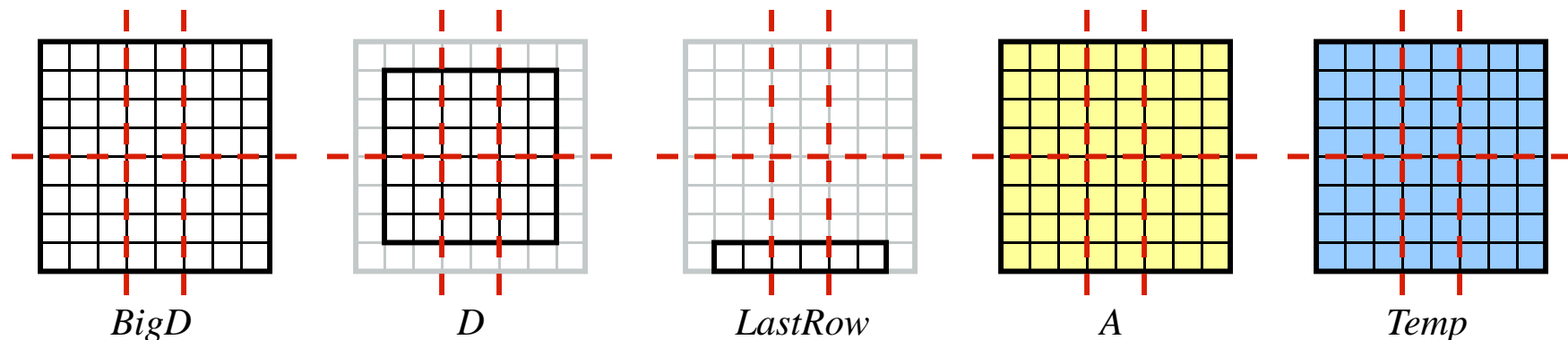
```

With this change, same code runs in a distributed manner

Domain distribution maps indices to *locales*

⇒ decomposition of arrays & default location of iterations over locales

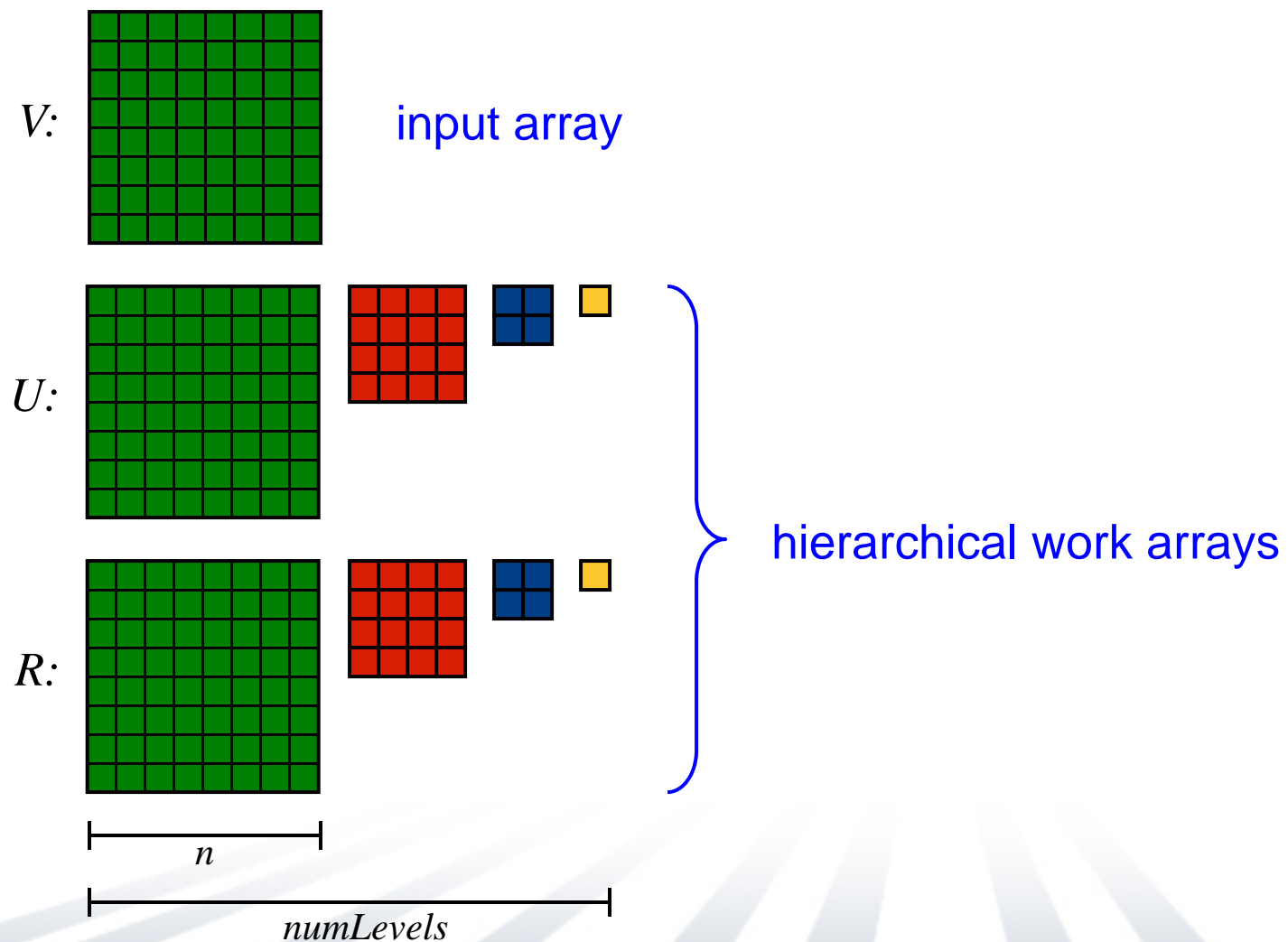
Subdomains inherit parent domain's distribution



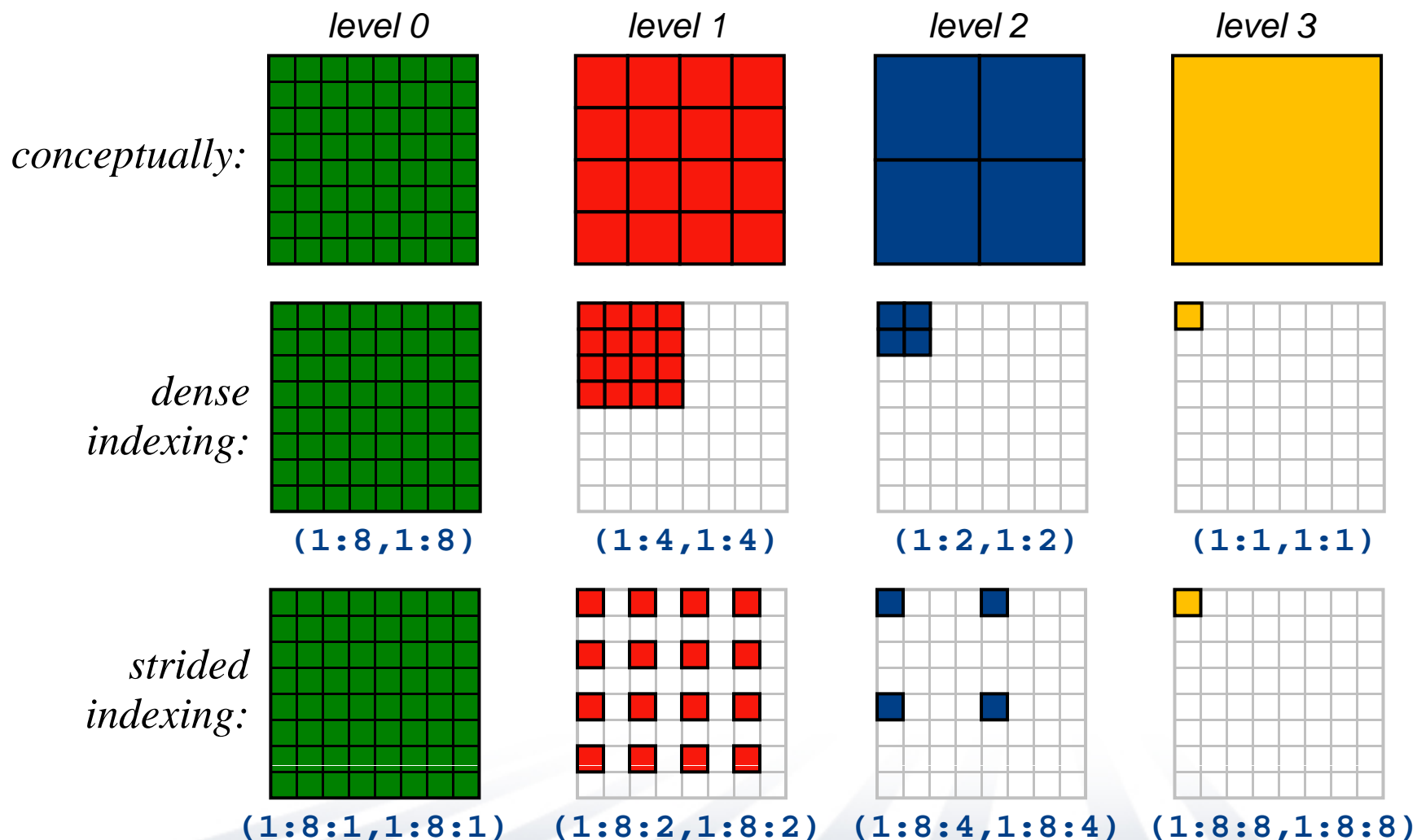
Jacobi Iteration in Chapel

```
config const n = 6,  
            epsilon = 1.0e-5;  
  
const BigD: domain(2) = [0..n+1, 0..n+1] distributed (Block),  
      D: subdomain(BigD) = [1..n, 1..n],  
      LastRow: subdomain(BigD) = D.exterior(1,0);  
  
var A, Temp : [BigD] real;  
  
A[LastRow] = 1.0;  
  
do {  
  [(i,j) in D] Temp(i,j) = (A(i-1,j) + A(i+1,j)  
                           + A(i,j-1) + A(i,j+1)) / 4.0;  
  
  var delta = max reduce abs(A(D) - Temp(D));  
  [ij in D] A(ij) = Temp(ij);  
} while (delta > epsilon);  
  
writeln(A);
```

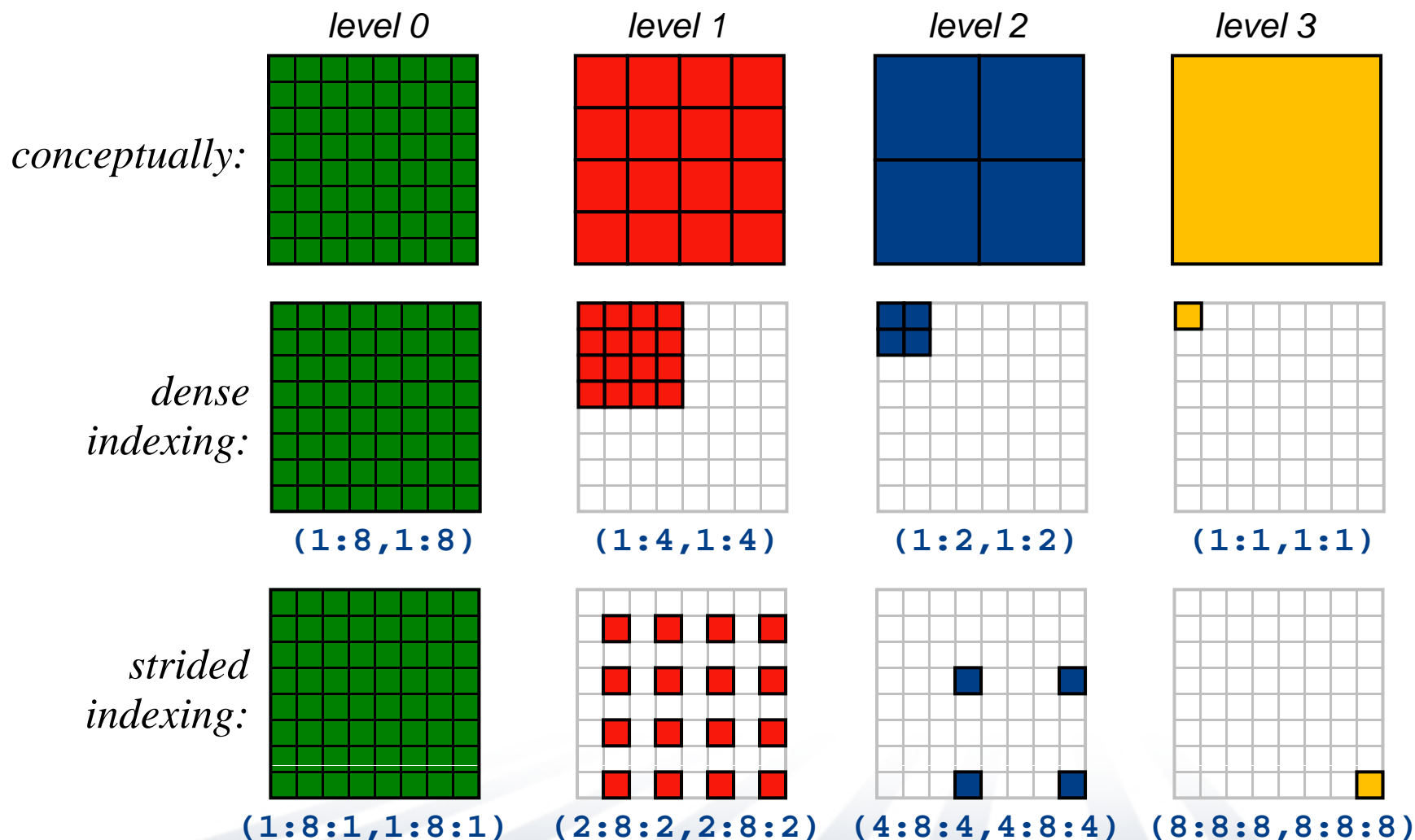
Example 2: Multigrid



Hierarchical Arrays



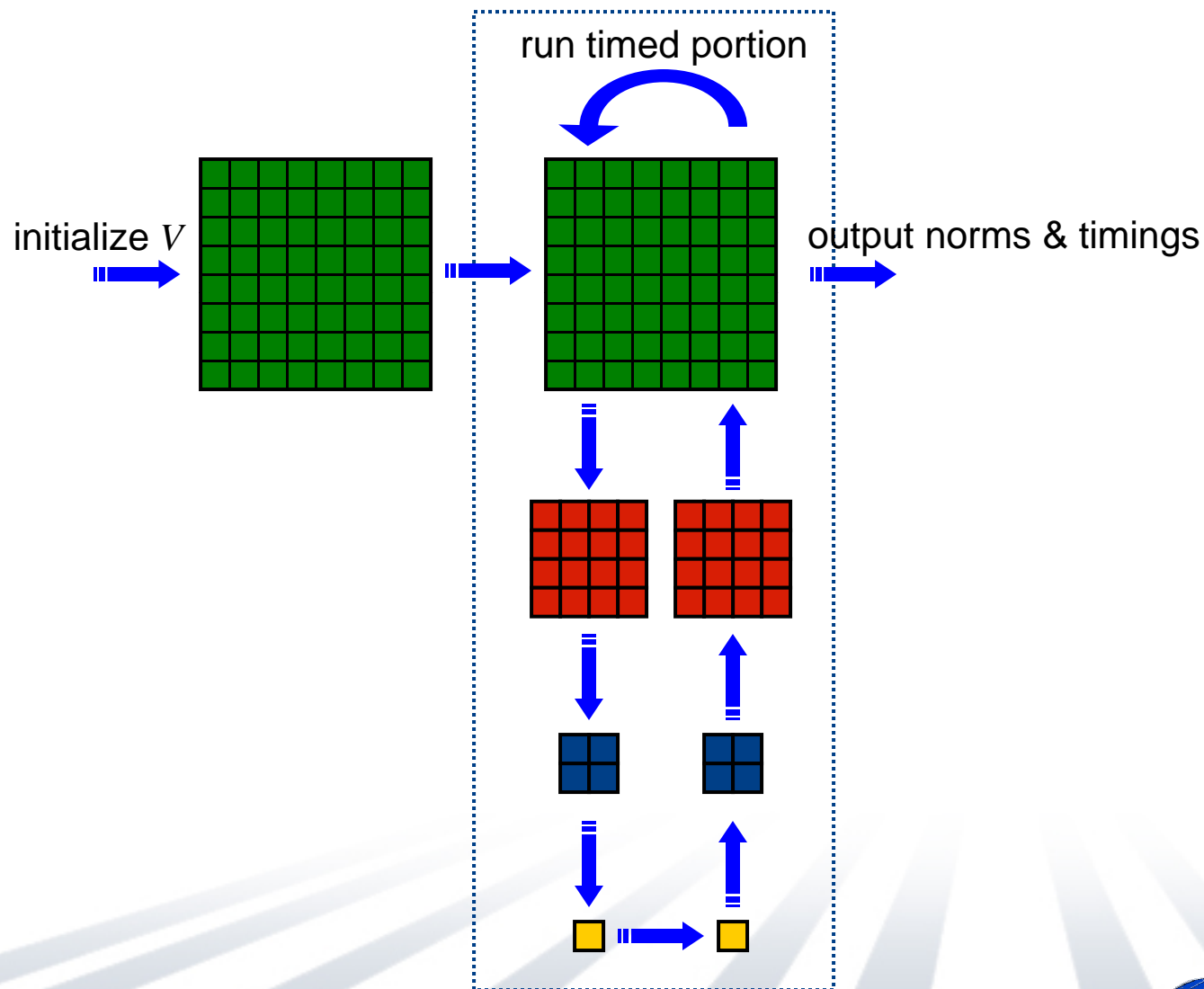
Hierarchical Arrays



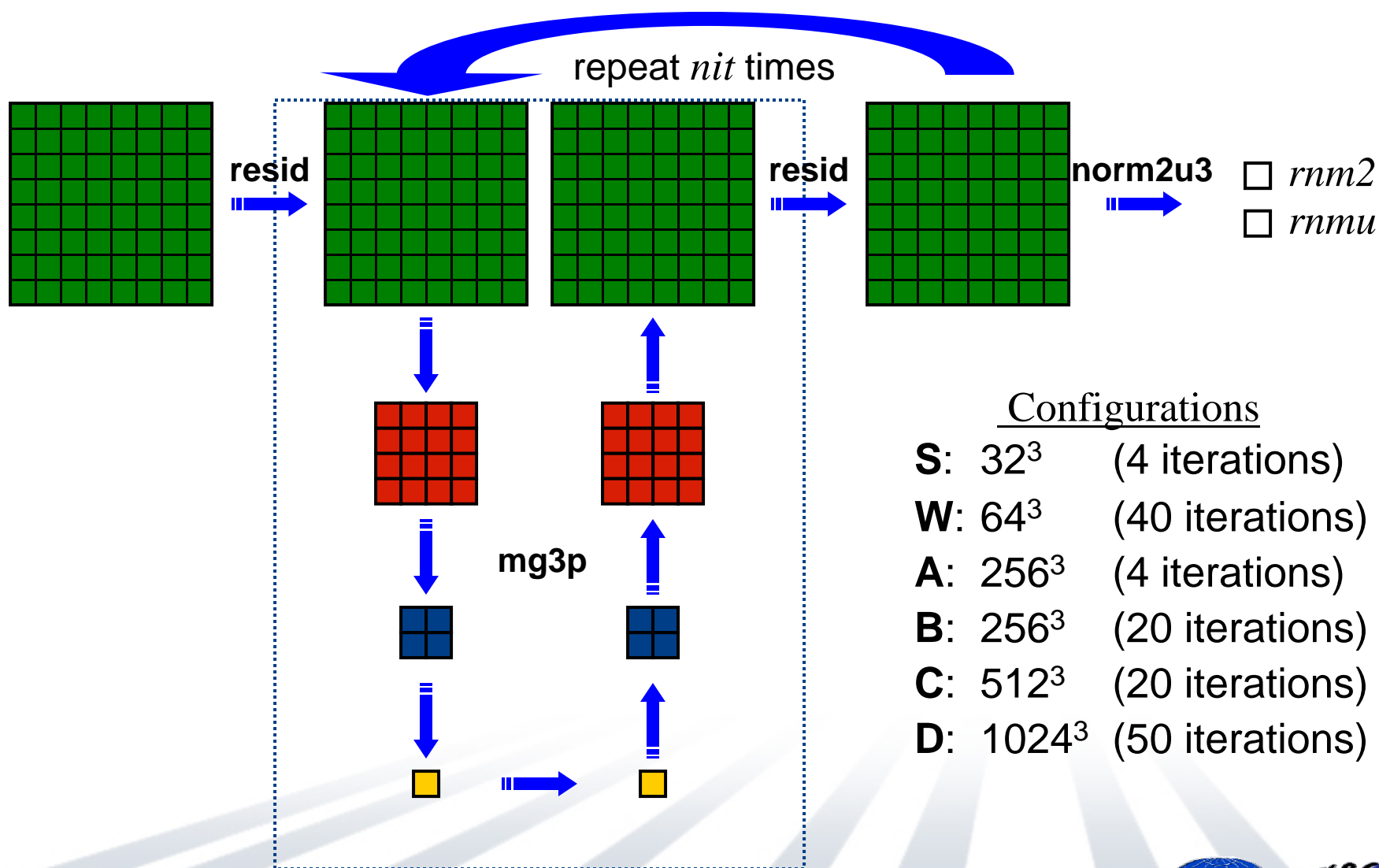
Hierarchical Array Declarations in Chapel

```
config const n = 1024,  
            numLevels = lg2(n);  
  
const Levels = [0..numLevels];  
const ProblemSpace: domain(1) distributed(Block) = (1..n)**3;  
  
var V: [ProblemSpace] real;  
  
const HierSpace: [lvl in Levels] subdomain(ProblemSpace)  
               = ProblemSpace by -2**lvl;  
  
var U, R: [lvl in Levels] [HierSpace(lvl)] real;
```

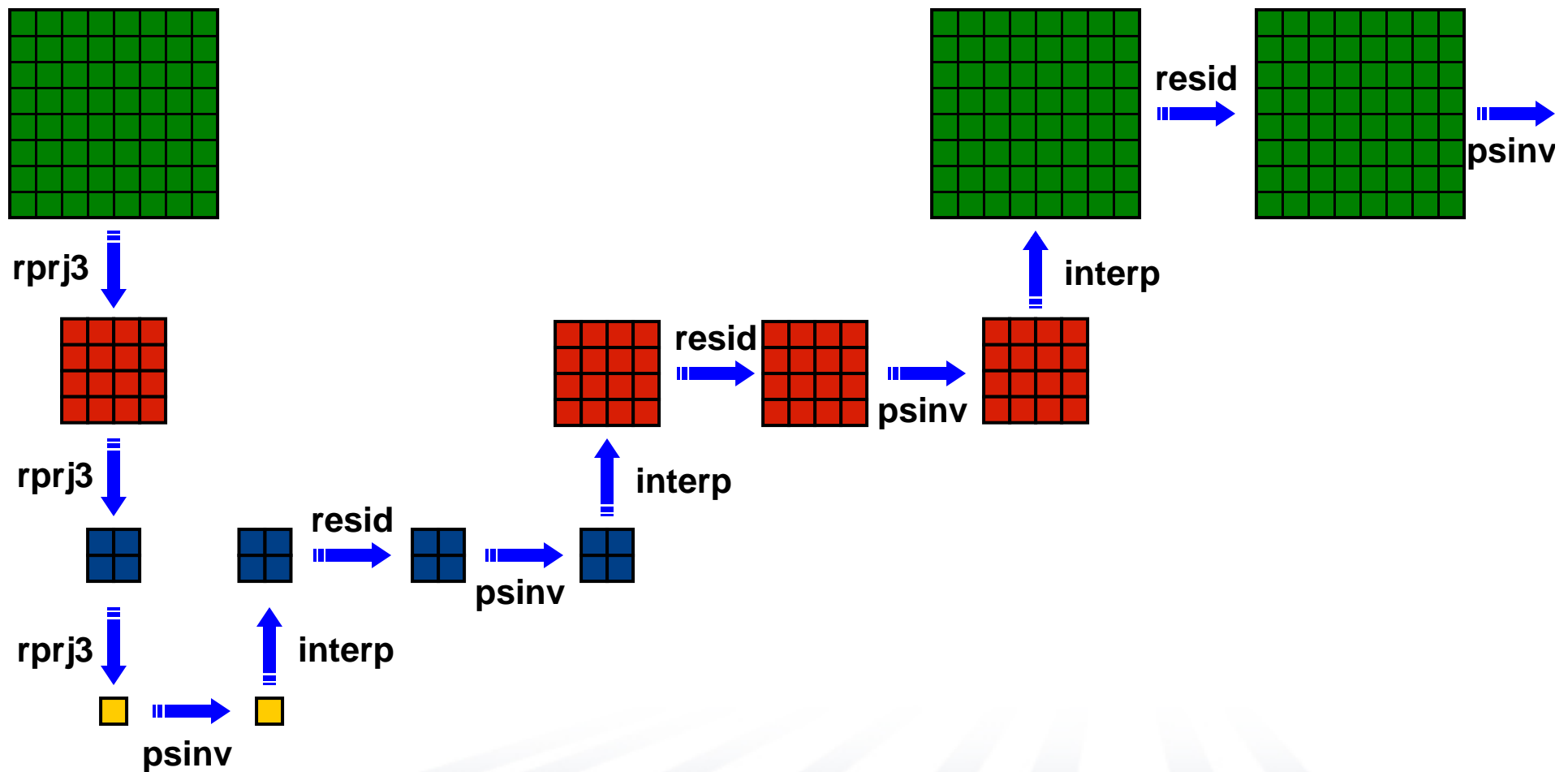
Overview of NAS MG



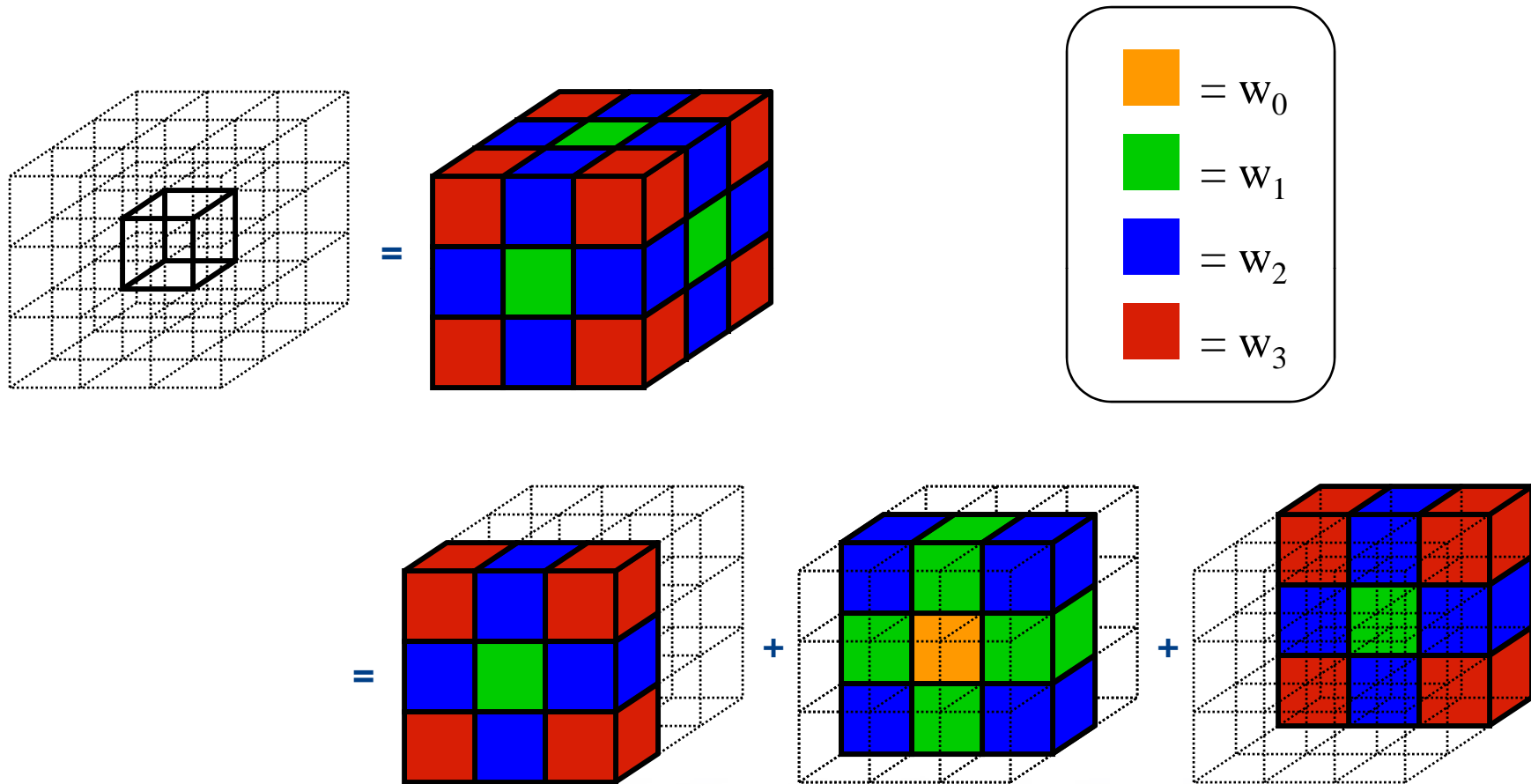
MG's Timed Portion



MG's projection/interpolation cycle



NAS MG: *rprj3* stencil



Multigrid: Stencils in Chapel

- Can write them out explicitly, as in Jacobi...

```
def rprj3(S, R) {
  param w: [0..3] real = (0.5, 0.25, 0.125, 0.0625);
  const Rstr = R.stride;

  forall ijk in S.domain do
    S(ijk) = w(0) * R(ijk)
      + w(1) * (R(ijk+Rstr*(1,0,0)) + R(ijk+Rstr*(-1,0,0))
        + R(ijk+Rstr*(0,1,0)) + R(ijk+Rstr*(0,-1,0))
        + R(ijk+Rstr*(0,0,1)) + R(ijk+Rstr*(0,0,-1)))
      + w(2) * (R(ijk+Rstr*(1,1,0)) + R(ijk+Rstr*(1,-1,0))
        + R(ijk+Rstr*(-1,1,0)) + R(ijk+Rstr*(-1,-1,0))
        + R(ijk+Rstr*(1,0,1)) + R(ijk+Rstr*(1,0,-1))
        + R(ijk+Rstr*(-1,0,1)) + R(ijk+Rstr*(-1,0,-1))
        + R(ijk+Rstr*(0,1,1)) + R(ijk+Rstr*(0,1,-1))
        + R(ijk+Rstr*(0,-1,1)) + R(ijk+Rstr*(0,-1,-1)))
      + w(3) * (R(ijk+Rstr*(1,1,1)) + R(ijk+Rstr*(1,1,-1))
        + R(ijk+Rstr*(1,-1,1)) + R(ijk+Rstr*(1,-1,-1))
        + R(ijk+Rstr*(-1,1,1)) + R(ijk+Rstr*(-1,1,-1))
        + R(ijk+Rstr*(-1,-1,1)) + R(ijk+Rstr*(-1,-1,-1)))
  }
```

Multigrid: Stencils in Chapel

- ...or, note that a stencil is simply a reduction over a small subarray expression
- Thus, stencils can be written in a “syntactically scalable” way using reductions:

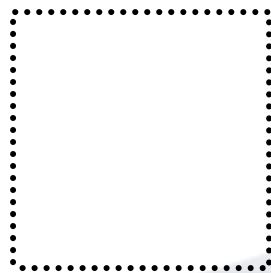
```
def rprj3(S, R) {  
    const Stencil: domain(3) = [-1..1, -1..1, -1..1], // 27-points  
    w: [0..3] real = (0.5, 0.25, 0.125, 0.0625), // 4 wgts  
    w3d = [(i,j,k) in Stencil] w((i!=0) + (j!=0) + (k!=0));  
  
    forall ijk in S.domain do  
        S(ijk) = + reduce [off in Stencil]  
                      (w3d(off) * R(ijk + R.stride*off));  
    }
```


Chapel (65)

Example 3: Fast Multipole Method (FMM)

```
var OSgfn, ISgfn: [lvl in Levels] [SpsCubes(lvl)] [Sgfn(lvl)] [1..3] complex;
```

1D array over levels
of the hierarchy



OSgfn(1)



OSgfn(2)



OSgfn(3)

Example 3: Fast Multipole Method (FMM)

```
var OSgfn, ISgfn: [lvl in Levels] [SpsCubes(lvl)] [Sgfns(lvl)] [1..3] complex;
```

1D array over levels
of the hierarchy

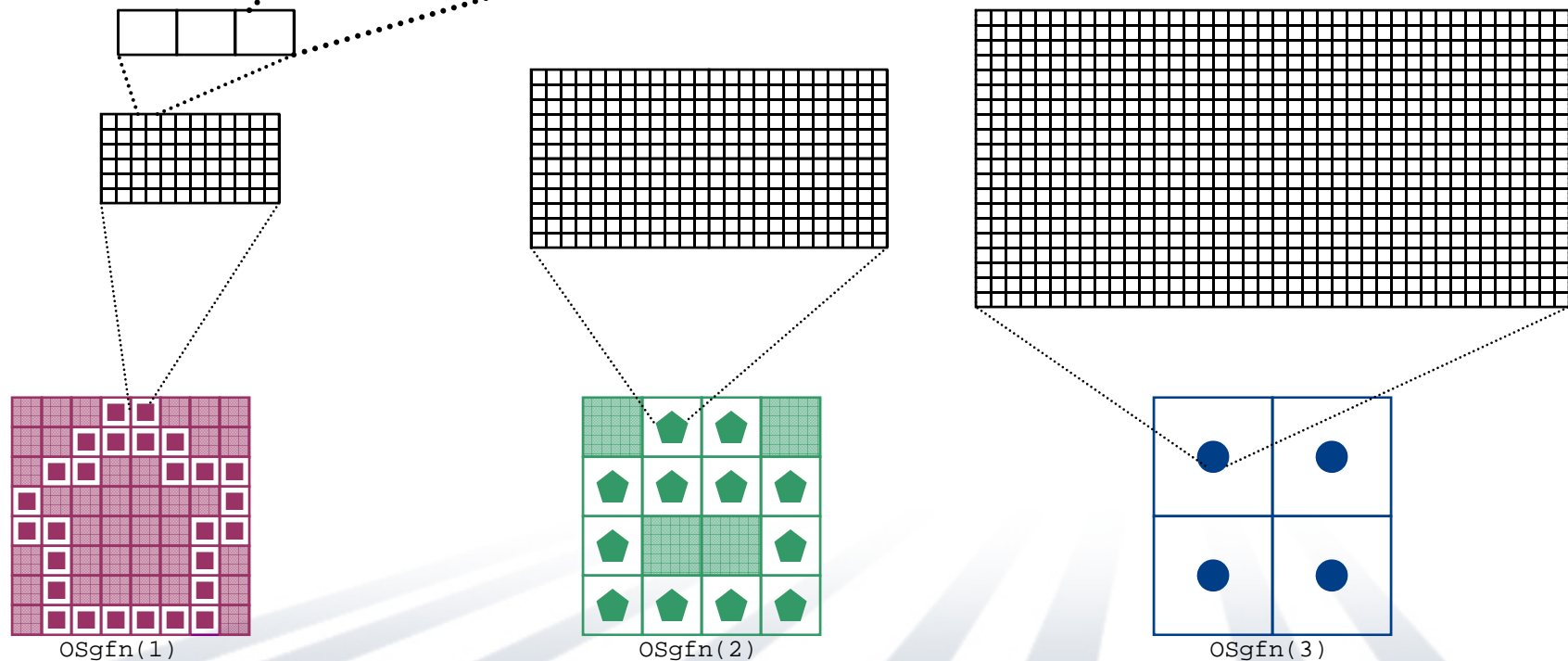
...of 3D sparse
arrays of cubes
(per level)

...of 1D vectors

...of 2D discretizations
of spherical functions,
(sized by level)

...of
complex
values

$$x + y \cdot i$$



FMM: Supporting Declarations

```
var OSgfn, ISgfn: [lvl in Levels] [SpsCubes(lvl)] [Sgfn(lvl)] [1..3] complex;
```

previous definitions:

```
var n: int = ...;
```

```
var numLevels: int = ...;
```

```
var Levels: domain(1) = [1..numLevels];
```

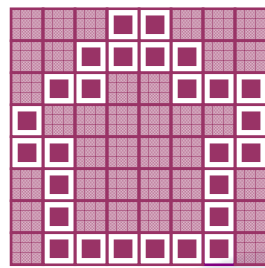
```
var scale: [lvl in Levels] int = 2** (lvl-1);
```

```
var SgFnSize: [lvl in Levels] int = computeSgFnSize(lvl);
```

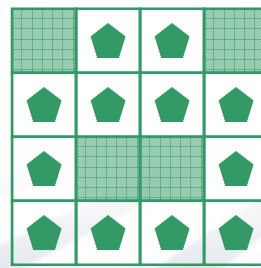
```
var LevelBox: [lvl in Levels] domain(3) = [(1,1,1)..(n,n,n)] by scale(lvl);
```

```
var SpsCubes: [lvl in Levels] sparse subdomain(LevelBox) = ...;
```

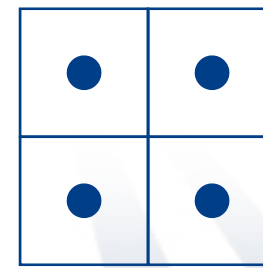
```
var Sgfn: [lvl in Levels] domain(2) = [1..SgFnSize(lvl), 1..2*SgFnSize(lvl)];
```



OSgfn(1)



OSgfn(2)



OSgfn(3)

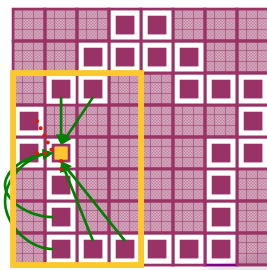
FMM: Computation

```
var OSgfn, ISgfn: [lvl in Levels] [SpsCubes(lvl)] [Sgfns(lvl)] [1..3] complex;
```

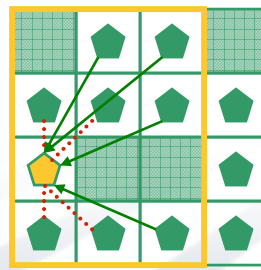
outer-to-inner translation:

```
for lvl in [1..numLevels) by -1 {
  ...
  forall cube in SpsCubes(lvl) {
    forall sib in out2inSiblings(lvl, cube) {
      const Trans = lookupXlateTab(cube, sib);

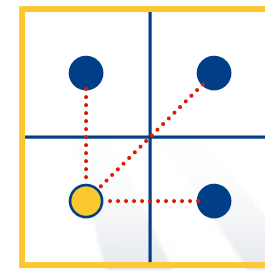
      atomic ISgfn(lvl)(cube) += OSgfn(lvl)(sib) * Trans;
    }
  }
  ...
}
```



OSgfn(1)



OSgfn(2)



OSgfn(3)

Fast Multipole Method: Summary

- Chapel code captures structure of data and computation far better than sequential Fortran/C versions (let alone MPI versions of them)
 - cleaner, more succinct, more informative
 - rich domain/array support plays a big role in this
- Code very clear to Boeing engineer familiar with FMM, unfamiliar with Chapel
- Parallelism shifts at different levels of hierarchy
 - Global view and syntactic separation of concerns helps here
 - Imagine writing in a fragmented language
- Yet, I've elided some non-trivial code (data distribution)

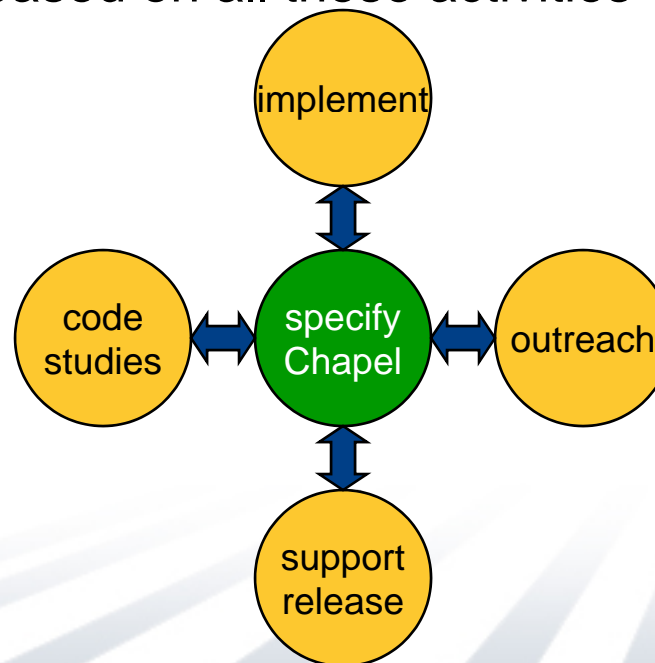
Outline

- ✓ Chapel Context
- ✓ Global-view Programming Models
- ✓ Language Overview
- ✓ Example Computations
- Status, Future Work, Collaborations

Chapel Work

■ Chapel Team's Focus:

- specify Chapel syntax and semantics
- implement prototype compiler for Chapel
- code studies of benchmarks, applications, and libraries in Chapel
- community outreach to inform and learn from users/researchers
- support users of preliminary releases
- refine language based on all these activities



Prototype Implementation

- Approach:
 - source-to-source compiler for portability (Chapel-to-C)
 - link against runtime libraries to hide machine details
 - threading layer currently implemented using pthreads
 - communication currently implemented using Berkeley's GASNet
- Status:
 - **base language:** solid, usable (a few gaps remain)
 - **task parallel:** multiple threads, multiple locales
 - **data parallel:** single-threaded, single-locale
 - **performance:** has received little effort (but much planning)
- Current Focus:
 - multi-threaded implementation of data parallel features
 - distributed domains and arrays
 - performance optimizations
 - hope to unveil first performance results at SC08 in Austin this fall
- Early releases to ~40 users at ~20 sites (academic, gov't, industry)

Research Challenges

■ Near-term:

- user-defined distributions
- zippered parallel iteration
- index/subdomain optimizations

■ Medium-term:

- memory management policies/mechanisms
- task scheduling policies
- tuning for multicore processors
- unstructured/graph-based codes
- compiling/optimizing atomic sections (STM)
- language interoperability
- parallel I/O

■ Longer-term:

- checkpoint/resiliency mechanisms
- exotic architectures (GPUs, FPGAs?)
- hierarchical/heterogeneous notion of locales
- increased static safety via type system

Chapel Design Philosophies

- A research project...
 - ...but intentionally broader than an academic project would tend to be
 - due to emphasis on general parallel programming
 - due to the belief that success requires a broad feature set
 - to create a platform for broad community involvement
- Nurture within Cray, then turn over to community
 - currently releasing to small set of “friendly” users
 - hope to do public release in late 2008
 - turn over to community when it can stand on its own

Collaborations

UIUC (Vikram Adve and Rob Bocchino): Software Transactional Memory (STM) over distributed memory (PPoPP `08)

ORNL (David Bernholdt *et al.*): Chapel code studies – Fock matrix computations, MADNESS, Sweep3D, ... (HIPS `08)

PNNL (Jarek Nieplocha *et al.*): ARMCI port of comm. layer

CMU (Franz Franchetti): Chapel as portable parallel back-end language for SPIRAL

EPCC (Michele Weiland, Thom Haddow): performance study of single-locale task parallelism

(Your name here?)

Possible Collaboration Areas

- any of the previously-mentioned research topics...
- task parallel concepts
 - implementation using alternate threading packages
 - work-stealing task implementation
- application/benchmark studies
- different back-ends (LLVM? MS CLR?)
- visualizations, algorithm animations
- library support
- tools
 - correctness debugging
 - performance debugging
 - IDE support
- runtime compilation
- (your ideas here...)

Chapel Team



Steve Deitz, Brad Chamberlain
David Iten, Samuel Figueroa, ~~Mary Beth Hribar~~

Questions?

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chapel_info@cray.com

<http://chapel.cs.washington.edu>

ZPL Sidebar



ZPL

ZPL: an array-based data parallel language

Developed by: University of Washington

Timeframe: 1991 – 2003 (can still download today)

Target machines: 1990's HPC parallel platforms

- clusters of commodity processors
- clusters of SMPs
- custom parallel architectures
 - Cray T3E, KSR, SGI Origin, IBM SP2, Sun Enterprise, ...

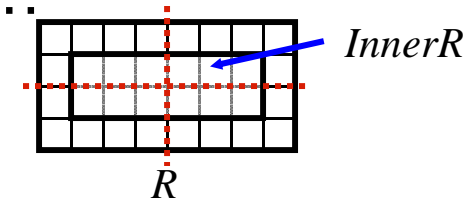
Main concepts:

- abstract machine model: CTA
- regions: first-class index sets
- WYSIWYG performance model

ZPL Concepts: Regions

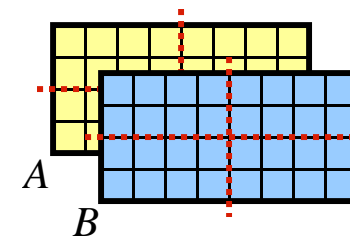
regions: first-class distributed index sets...

```
region R      = [1..m, 1..n];
    InnerR = [2..m-1, 2..n-1];
```



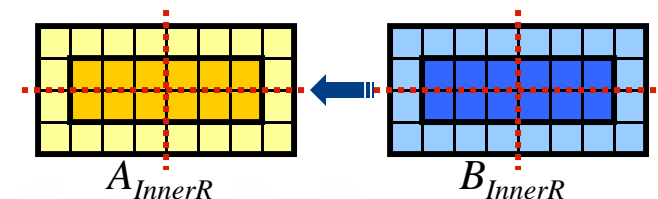
...used to declare distributed arrays...

```
var A, B: [R] double;
```



...and computation over distributed arrays

```
[InnerR] A = B;
```



ZPL Concepts: Array Operators

array operators: describe nontrivial array indexing

at operator (@): translation

[InnerR] A = B@[0,1];

flood operator (>>): replication

[R] A = >>[1, 1..n] B;

reduction operator (op<<): reductions

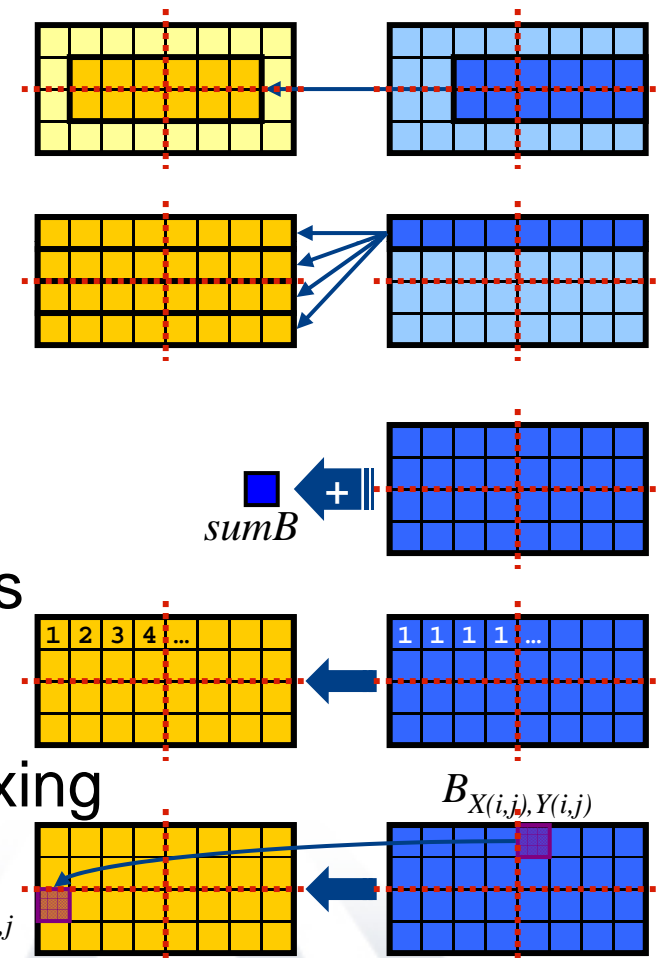
[R] sumB = +<< B;

scan operator (op||): parallel prefixes

[R] A = +|| B;

remap operator (#): whole-array indexing

[R] A = B#[X,Y];



ZPL Concepts: Syntactic Performance Model

[InnerR] A = B;

No Array Operators \Rightarrow
No Communication

[InnerR] A = B@[0,1];

At Operator \Rightarrow
Point-to-Point Communication

[R] A = >>[1, 1..n] B;

Flood Operator \Rightarrow Broadcast
(log-tree) Communication

[R] sumB = +<< B;

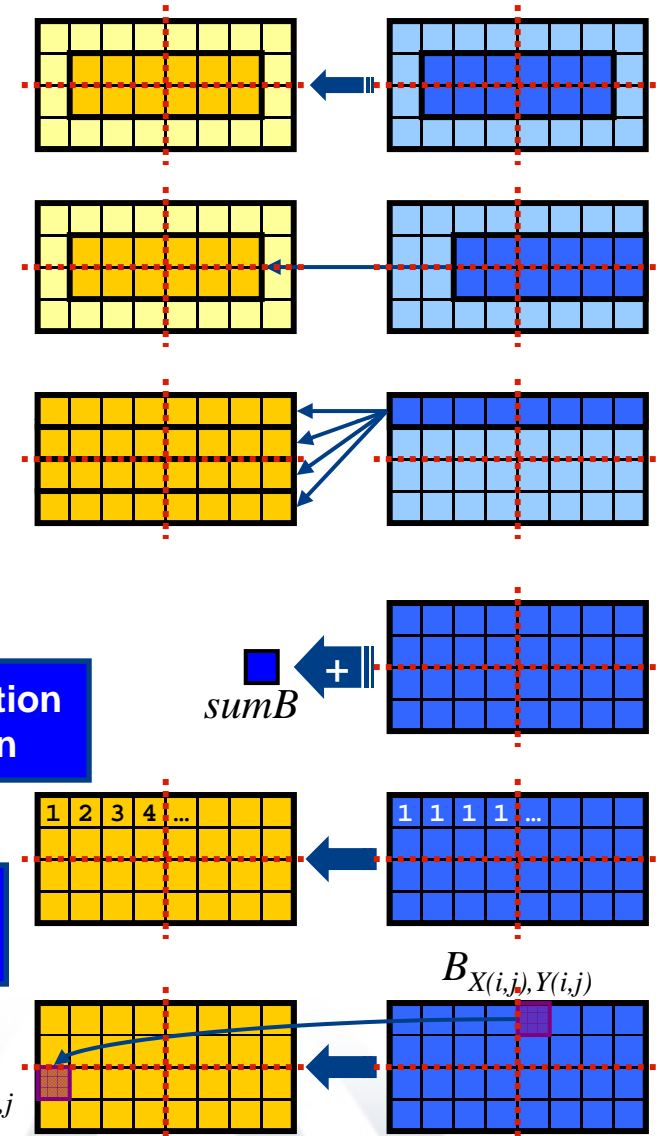
Reduce Operator \Rightarrow Reduction
(log-tree) Communication

[R] A = +|| B;

Scan Operator \Rightarrow Parallel-Prefix
(log-tree) Communication

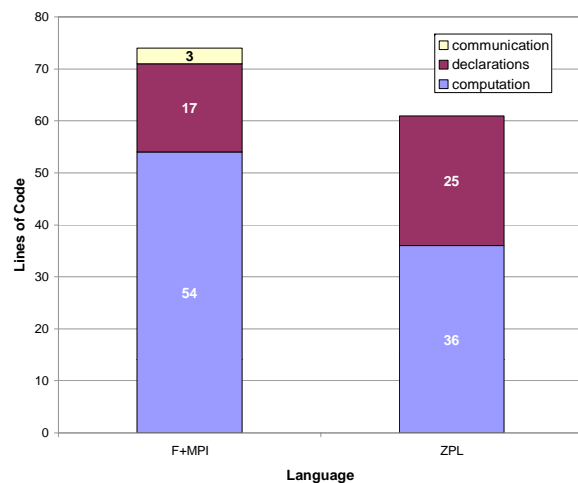
[R] A = B#[X,Y];

Remap Operator \Rightarrow Arbitrary
(all-to-all) Communication

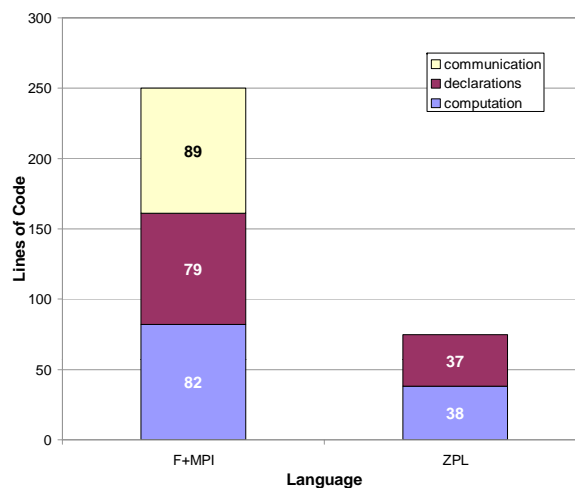


NPB: MPI vs. ZPL Code Size (timed kernels)

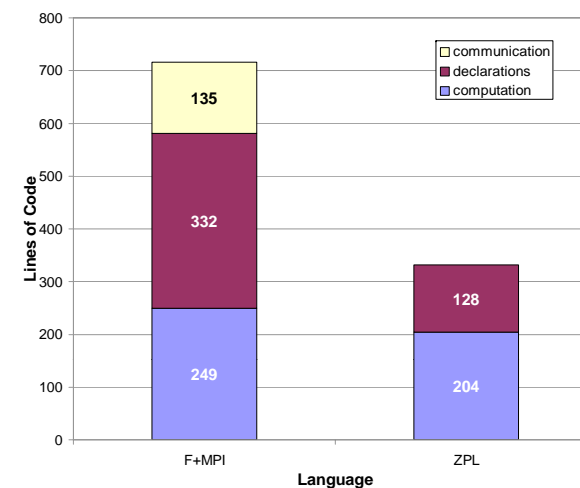
EP



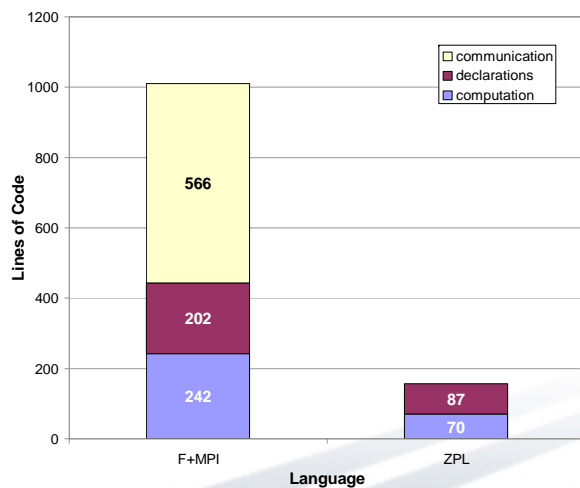
CG



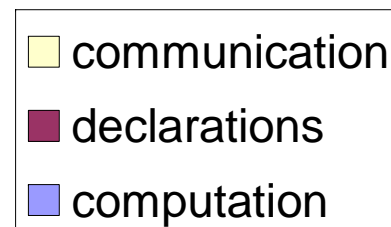
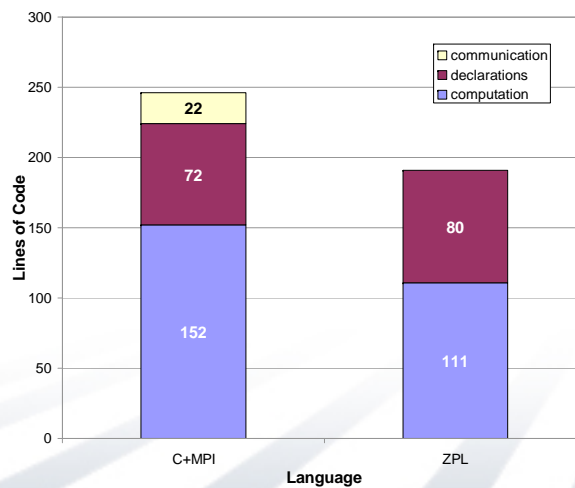
FT



MG

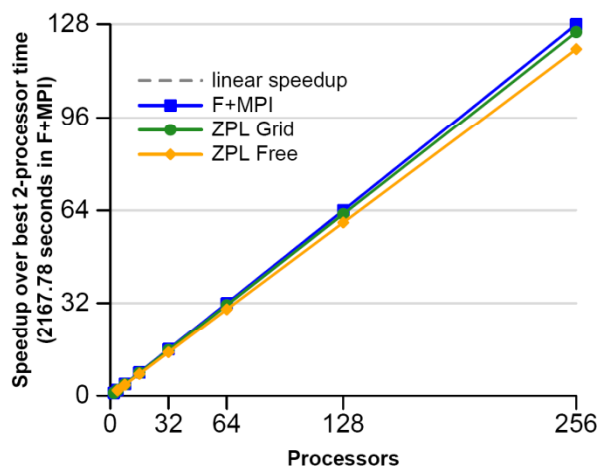


IS

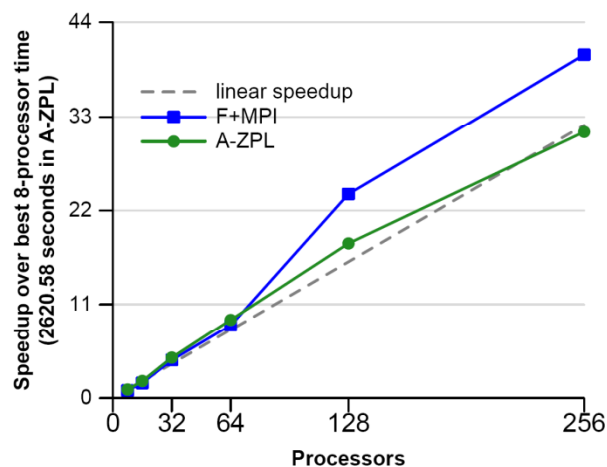


NPB: MPI vs. ZPL Performance

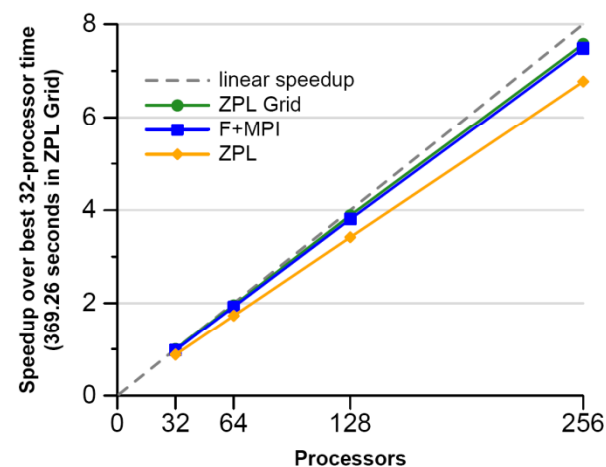
EP



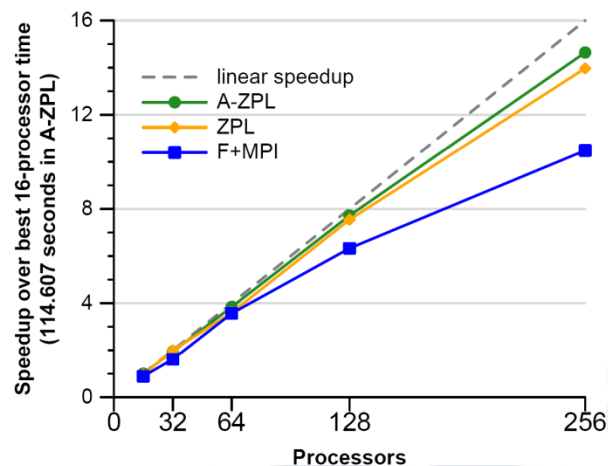
CG



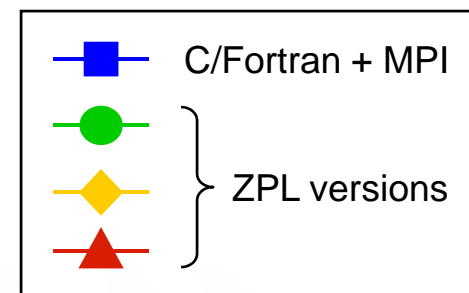
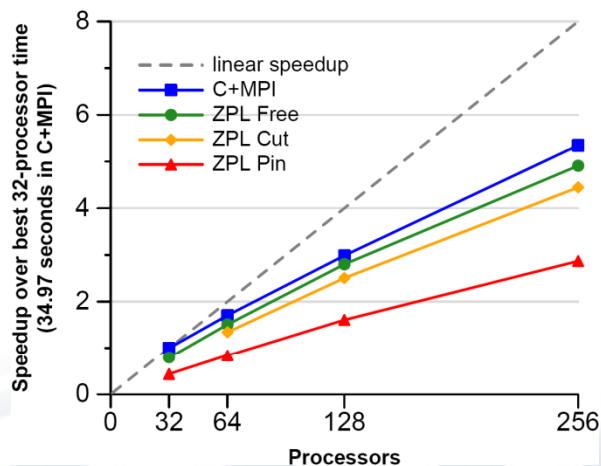
FT



MG



IS



ZPL Summary

- + Global-view programming with syntactic performance model
 - good for the compiler
 - good for the performance-oriented user
- + concise/clean compared to MPI, w/ competitive performance
- only supports a single level of data parallelism
- only supports a small set of distributions
- distinct concepts for sequential and parallel arrays

For more information:

<http://cs.washington.edu/research/zpl>

zpl-info@cs.washington.edu

HoPL'07 paper about ZPL

ZPL

ZPL strengths

- + syntactic performance model (e.g., communication visible in source)
 - helps user reason about program's parallel implementation
 - helps compiler implement and optimize it
- + global view of data and computation
 - programmer need not think in SPMD
- + implementation-neutral expression of communication
 - permits mapping to best mechanisms for given architecture/level

ZPL weaknesses

- only supports one level of data parallelism; no true task parallelism
 - a consequence of its use of an SPMD execution model
- distinct concepts for parallel and serial arrays
- only supports a small number of built-in distributions

But let's take the lessons from ZPL that we can and keep striving forward... (and from other “failed” 1990's parallel languages as well)

(RETURN)



NAS MG *rprj3* stencil in ZPL

```

procedure rprj3(var S,R: [,] double;
                d: array [] of direction);
begin
    S := 0.5      * R
      + 0.25      * (R@^d[ 1, 0, 0] + R@^d[ 0, 1, 0] + R@^d[ 0, 0, 1] +
                    R@^d[-1, 0, 0] + R@^d[ 0,-1, 0] + R@^d[ 0, 0,-1])
      + 0.125     * (R@^d[ 1, 1, 0] + R@^d[ 1, 0, 1] + R@^d[ 0, 1, 1] +
                    R@^d[ 1,-1, 0] + R@^d[ 1, 0,-1] + R@^d[ 0, 1,-1] +
                    R@^d[-1, 1, 0] + R@^d[-1, 0, 1] + R@^d[ 0,-1, 1] +
                    R@^d[-1,-1, 0] + R@^d[-1, 0,-1] + R@^d[ 0,-1,-1])
      + 0.0625    * (R@^d[ 1, 1, 1] + R@^d[ 1, 1,-1] +
                    R@^d[ 1,-1, 1] + R@^d[ 1,-1,-1] +
                    R@^d[-1, 1, 1] + R@^d[-1, 1,-1] +
                    R@^d[-1,-1, 1] + R@^d[-1,-1,-1]);
end;

```

Chapel (90)

```

if( axis .eq. 2 )then
  do i3=2,n3-1
    do ii=1,n1
      indx = indx + 1
      u(ii,i3) = buff(indx, buff_id )
    enddo
  enddo
endif

if( axis .eq. 3 )then
  do i2=1,n2
    do ii=1,n1
      indx = indx + 1
      u(ii,i2,1) = buff(indx, buff_id )
    enddo
  enddo
endif

return
end

subroutine rpej3(r,mlk,m2k,m3k,s,m1j,m2j,m3j,k)
implicit none
include 'cafnpb.h'
include 'globals.h'

integer mlk, m2k, m3k, m1j, m2j, m3j, k

double precision r(mlk,m2k,m3k), s(m1j,m2j,m3j)
integer j3, j2, j1, i3, i2, i1, d1, d2, d3, j

double precision x1(m), y1(m), x2,y2

if(mlk.eq.3)then
  d1 = 2
else
  d1 = 1
endif

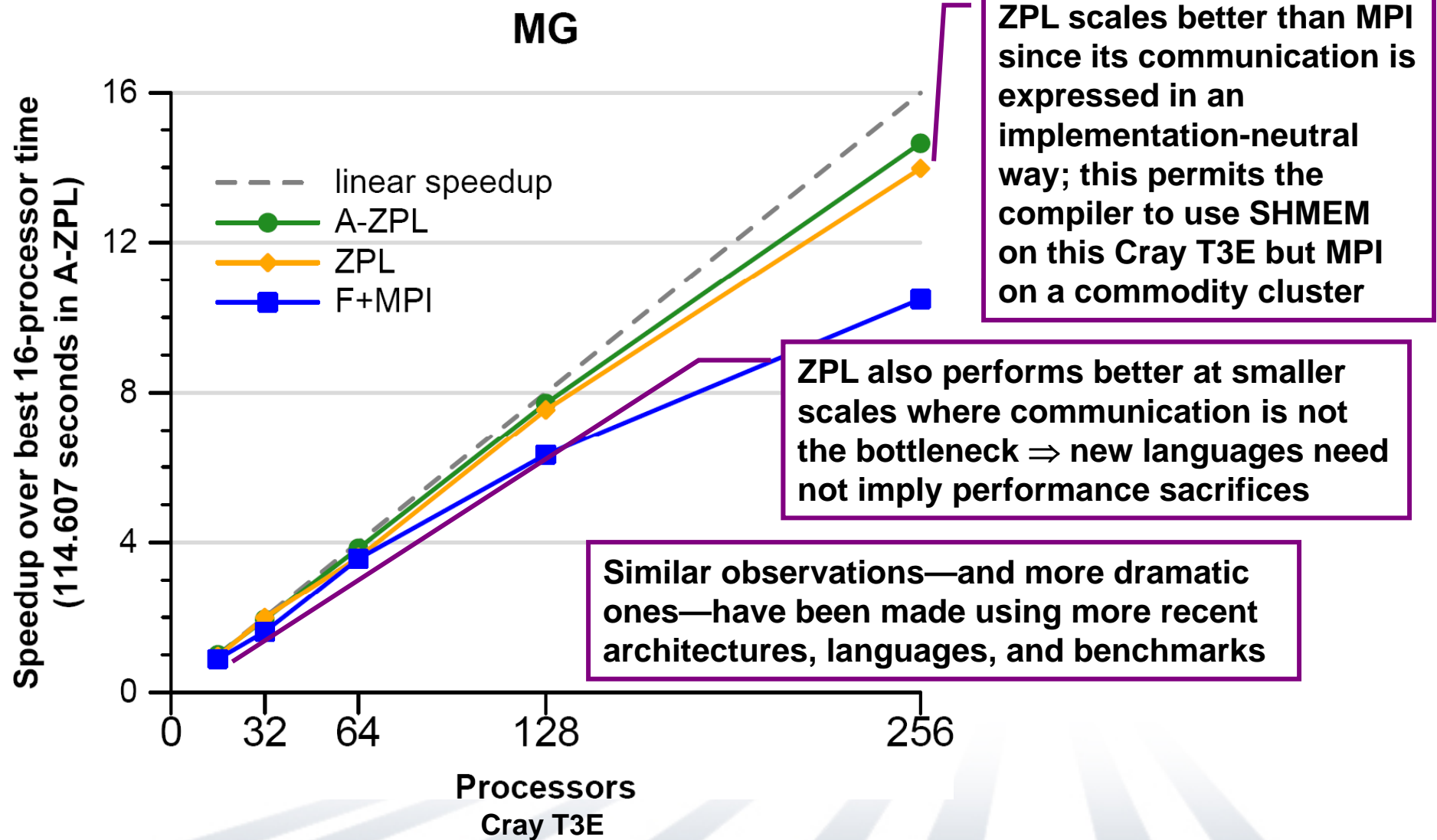
if(m2k.eq.3)then
  d2 = 2
else
  d2 = 1
endif

if(m3k.eq.3)then
  d3 = 2
else
  d3 = 1
endif

do j3=2,m3j-1
  i3 = 2*j3-d3
  do j2=2,m2j-1
    i2 = 2*j2-d2
    do j1=2,m1j
      i1 = 2*j1-d1
      x1(i1-1) = r(i1-1,i2-1,i3) + r(i1-1,i2+1,i3)
>      + r(i1-1,i2,i3-1) + r(i1-1,i2,i3+1)
>      + r(i1-1,i2+1,i3-1) + r(i1-1,i2+1,i3+1)
      enddo
      do j1=2,m1j-1
        i1 = 2*j1-d1
        y2 = r(i1,i2-1,i3-1) + r(i1,i2-1,i3+1)
>      + r(i1,i2+1,i3-1) + r(i1,i2+1,i3+1)
>      + r(i1,i2-1,i3) + r(i1,i2+1,i3)
>      + r(i1,i2,i3-1) + r(i1,i2,i3+1)
        s(j1,j2,j3) =
>      + 0.5D0 * r(i1,i2,i3)
>      + 0.25D0 * ( r(i1-1,i2,i3) + r(i1+1,i2,i3) + x2)
>      + 0.125D0 * ( x1(i1-1) + x1(i1+1) + y2)
>      + 0.0625D0 * ( y1(i1-1) + y1(i1+1) )
      enddo
    enddo
  enddo
  j = k-1
  call comm3(s,m1j,m2j,m3j,j)
  return
end

```

Performance Notes

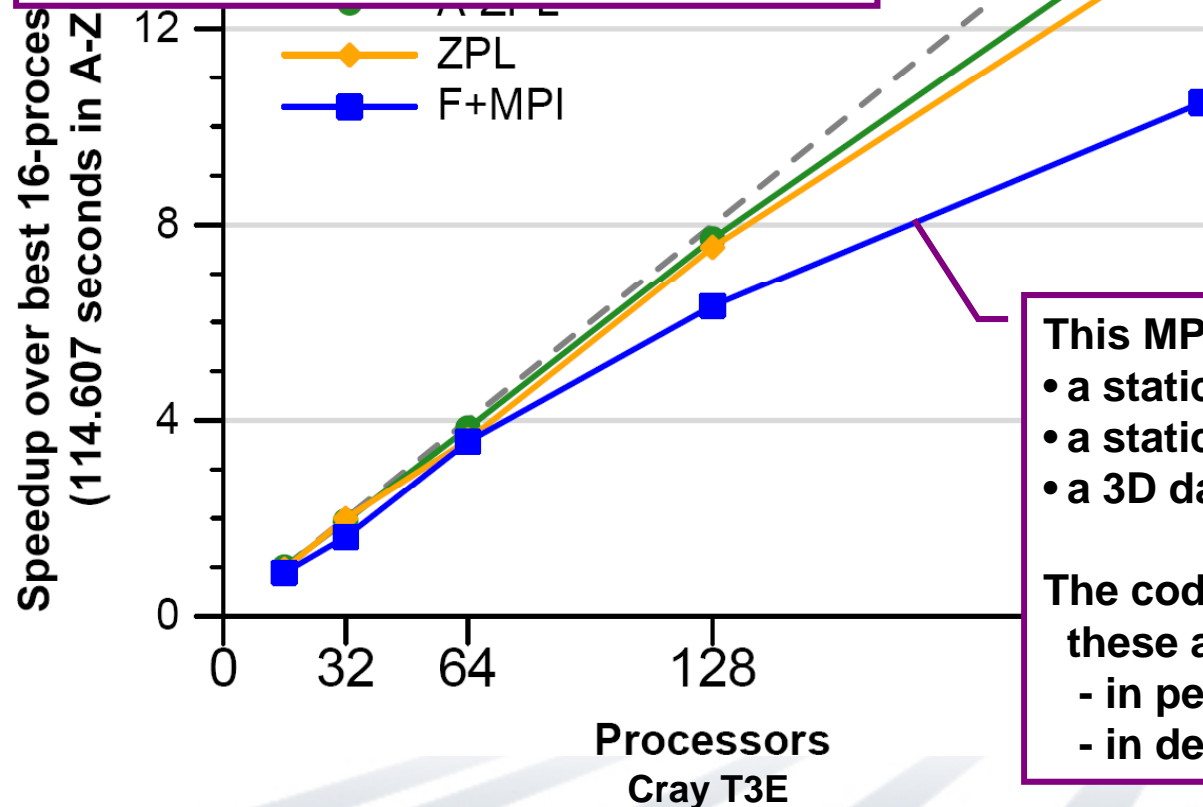


Generality Notes

MG

Each ZPL binary supports:

- an arbitrary load-time problem size
- an arbitrary load-time # of processors
- 1D/2D/3D data decompositions



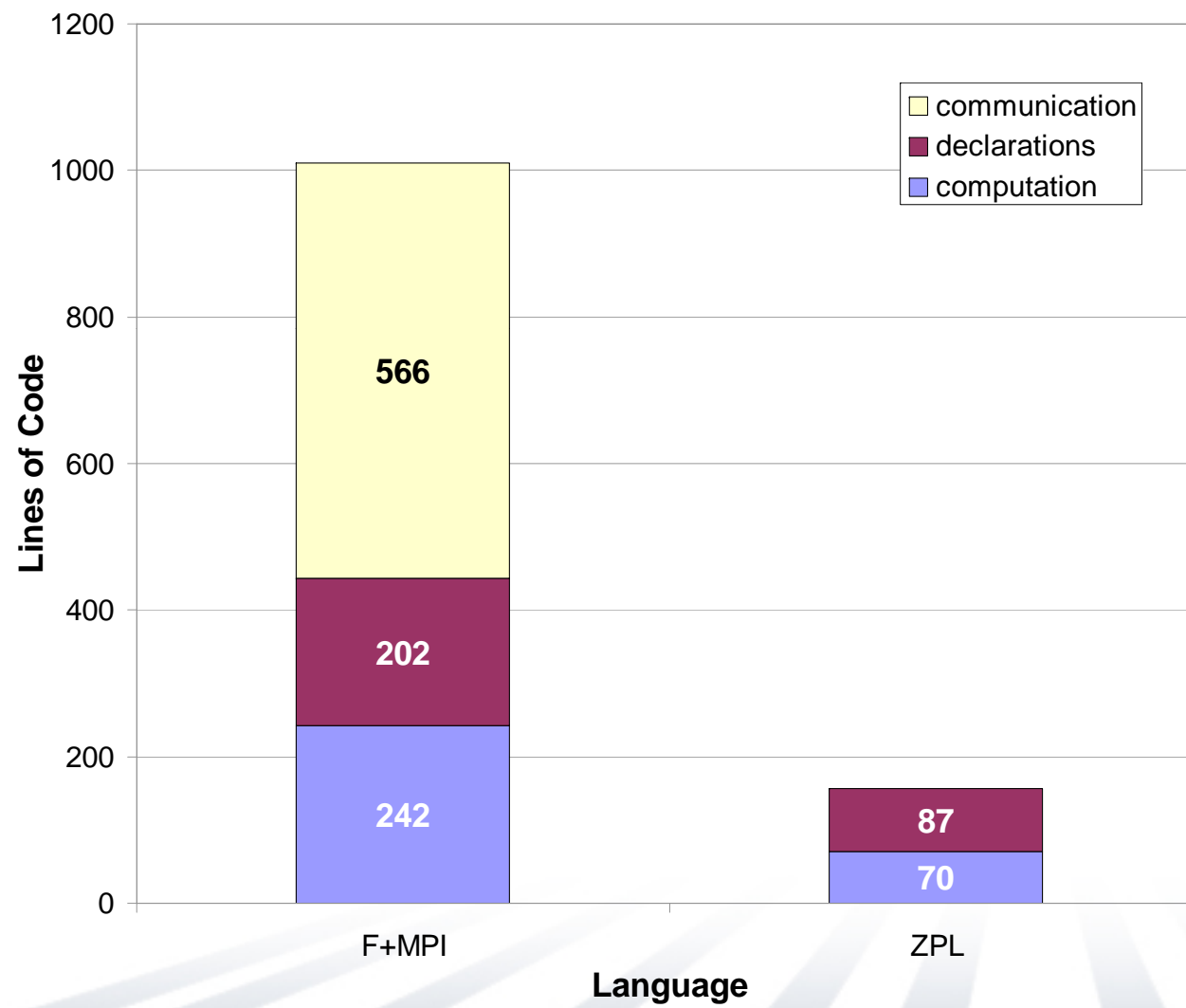
This MPI binary only supports:

- a static 2^k problem size
- a static 2^j # of processors
- a 3D data decomposition

The code could be rewritten to relax these assumptions, but at what cost?

- in performance?
- in development effort?

Code Size Notes



Code Size Notes

- the ZPL is 6.4x shorter because it supports finer-grain parallelism than the cooperating executable
- in particular, it's not an SPMD programming model
 - ⇒ little/no code for communication
 - ⇒ little/no code for array bookkeeping

More important than the size difference is that it is easier to write, read, modify, and maintain

