

Chapel

the Cascade High Productivity Language

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What is Chapel?

- A new parallel language being developed by Cray Inc.
- Part of Cray's entry in DARPA's HPCS program
- **Main Goal:** Improve programmer productivity
 - Improve the **programmability** of parallel computers
 - Match or beat the **performance** of current programming models
 - Provide better **portability** than current programming models
 - Improve **robustness** of parallel codes
- Target architectures:
 - multicore desktop machines
 - clusters of commodity processors
 - Cray architectures
 - systems from other vendors
- A work in progress

Chapel's Setting: HPCS

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

- **Goal:** Raise productivity of high-end computing users by 10×
- **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 –)
 - Implement the systems and technologies resulting from phase II
 - (Sun also continues work on Fortress, without HPCS funding)

Chapel: Motivating Themes

- 1) general parallel programming
- 2) *global-view* abstractions
- 3) *multiresolution* design
- 4) control of locality/affinity
- 5) reduce gap between mainstream & parallel languages

Outline

- ✓ Chapel Context
- Chapel Themes
- Language Overview
- Status, Collaborations, Future Work

1) General Parallel Programming

■ General software parallelism

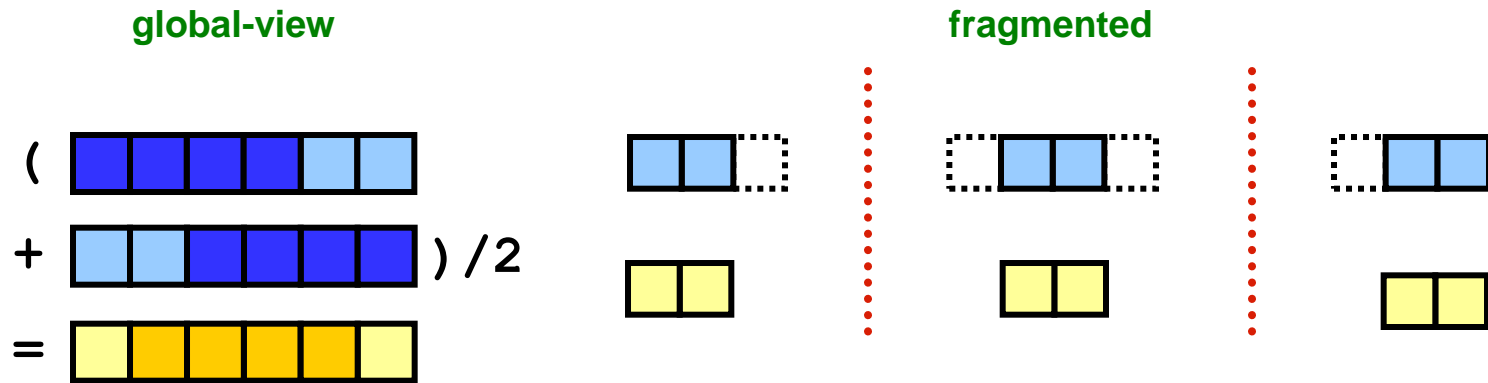
- *Algorithms*: should be able to express any that come to mind
 - should never hit a limitation requiring the user to return to MPI
- *Styles*: data-parallel, task-parallel, concurrent algorithms
 - as well as the ability to compose these naturally
- *Levels*: module-level, function-level, loop-level, statement-level, ...

■ General hardware parallelism

- *Types*: multicore desktops, clusters, HPC systems, ...
- *Levels*: inter-machine, inter-node, inter-core, vectors, multithreading

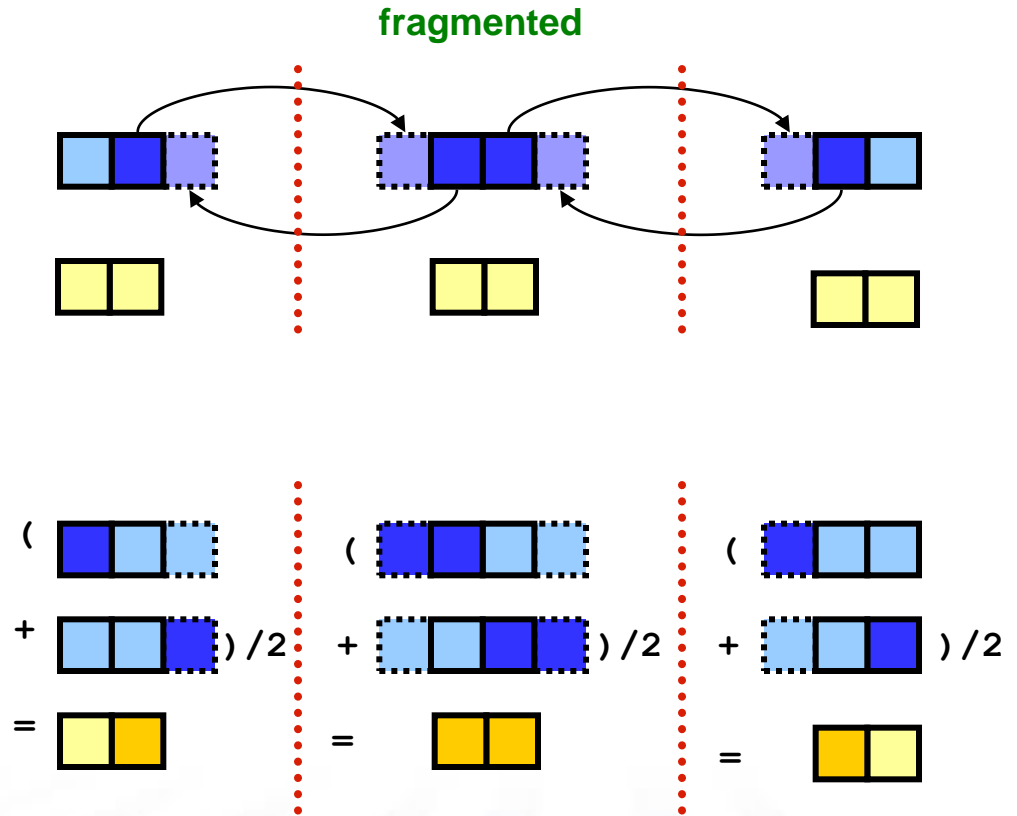
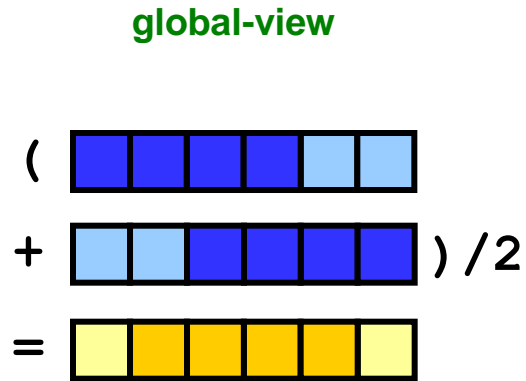
2) Global-view vs. Fragmented

Problem: “Apply 3-pt stencil to vector”



2) Global-view vs. Fragmented

Problem: “Apply 3-pt stencil to vector”



2) 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;
  }
}
```



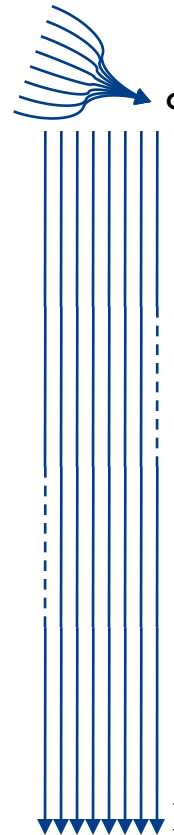
SPMD

```
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;
  }
}
```



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;
  }
}
```



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));
    rcv(right, a(locN+1));
  } else {
    innerHi = locN-1;
  }

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

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



2) SPMD pseudo-code + MPI

Problem: “Apply 3-pt stencil to vector”

SPMD (pseudocode + MPI)

```

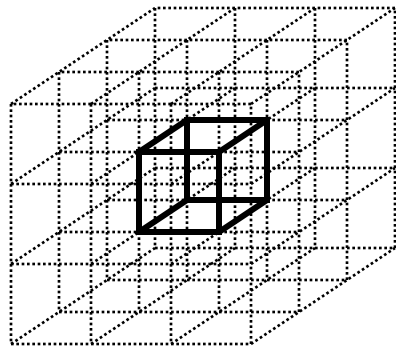
var n: int = 1000, locN: int = n/numProcs;
var a, b: [0..locN+1] real;
var innerLo: int = 1, innerHi: int = locN;
var numProcs, myPE: int;
var retval: int;
var status: MPI_Status;

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;
}

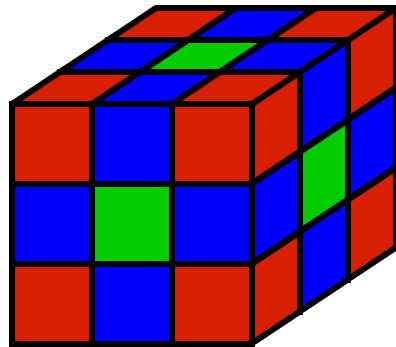
```





Communication becomes geometrically more complex for higher-dimensional arrays

2) *rprj3* stencil from NAS MG

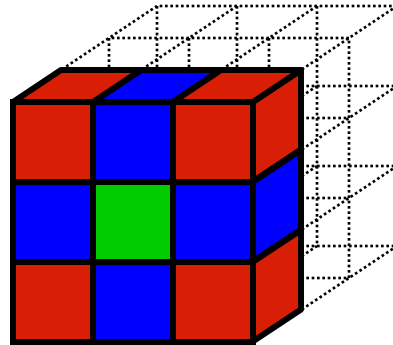


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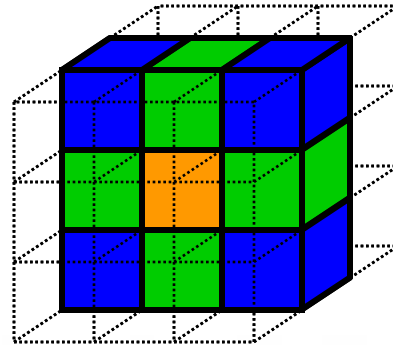


	= W_0
	= W_1
	= W_2
	= W_3

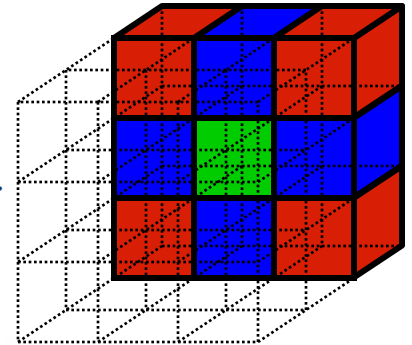
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+



+



2) NAS MG *rprj3* stencil in Fortran + MPI

```

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( nprocs .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 commp( 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)
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 i2=1,n2
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 i2=1,n2
do i1=1,n1
buff_len = buff_len + 1
buff(buff_len, buff_id) = u( i1, i2, n3-1 )
enddo
enddo
> buff(1:buff_len, buff_id+1) [nbr(axis, dir, k)] =
buff(1:buff_len, buff_id)
endif
endif
return
end

subroutine commlp( 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(i, buff_id) = 0.0D0
enddo

dir = +1
buff_id = 3 + dir
buff_len = nm2
do i=1, nm2
buff(i, 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
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)
1, i3 )
enddo
enddo
endif
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, n2-1, i3 )
enddo
enddo
> buff(1:buff_len, buff_id+1) [nbr(axis, dir, k)] =
buff(1:buff_len, buff_id)
endif
endif

subroutine take3( axis, dir, u, n1, n2, n3 )
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_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(i1, 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, i2, i3 )
enddo
enddo
> buff(1:buff_len, buff_id+1) [nbr(axis, dir, k)] =
buff(1:buff_len, buff_id)
1, i3 )
enddo
enddo
endif
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, n2-1, i3 )
enddo
enddo
> buff(1:buff_len, buff_id+1) [nbr(axis, dir, k)] =
buff(1:buff_len, buff_id)
endif
endif

subroutine rprj3(r,m1k,m2k,m3k,s,m1j,m2j,m3j,k)
implicit none
include 'cafnpb.h'
include 'globals.h'

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

double precision r(m1k,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(m1k.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)
> 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
enddo
enddo
j = k-1
call comm3(s,m1j,m2j,m3j,j)
return
end

```

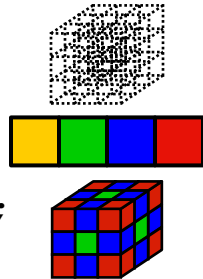
2) NAS MG *rprj3* stencil in Chapel

```

def rprj3(S, R) {
  const Stencil = [-1..1, -1..1, -1..1],
         w: [0..3] real = (0.5, 0.25, 0.125, 0.0625),
         w3d = [(i,j,k) in Stencil] w((i!=0) + (j!=0) + (k!=0));

  forall ijk in S.domain do
    S(ijk) = + reduce [offset in Stencil]
                (w3d(offset) * R(ijk + offset*R.stride));
}

```



Our previous work in ZPL showed that compact, global-view codes like these can result in performance that matches or beats hand-coded Fortran+MPI while also supporting more runtime flexibility

2) Classifying HPC Programming Notations

■ communication libraries:

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

data / control

fragmented / fragmented/SPMD
fragmented / SPMD

■ shared memory models:

- OpenMP, pthreads

global-view / global-view (trivially)

■ PGAS languages:

- Co-Array Fortran
- UPC
- Titanium

fragmented / SPMD
global-view / SPMD
fragmented / SPMD

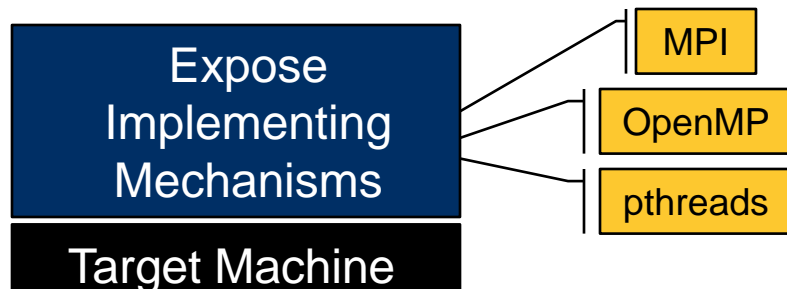
■ HPCS languages:

- Chapel
- X10 (IBM)
- Fortress (Sun)

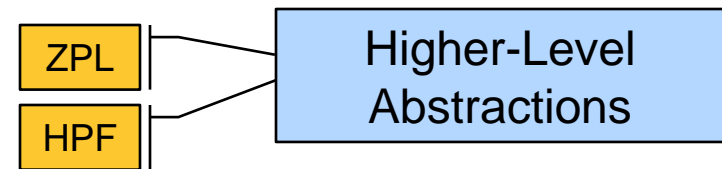
global-view / global-view
global-view / global-view
global-view / global-view

3) Multiresolution Languages: Motivation

Two typical camps of parallel language design:
 low-level vs. high-level



“Why is everything so tedious?”

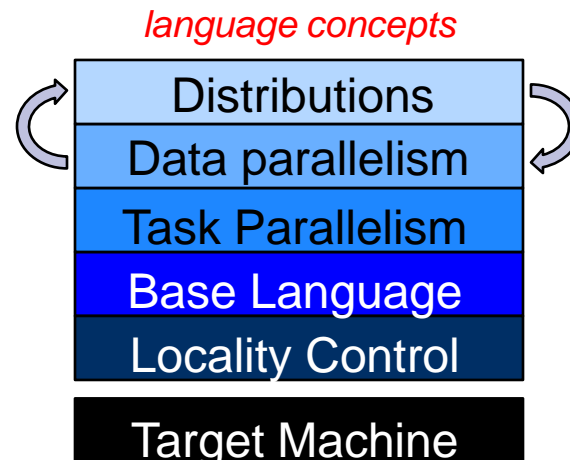


“Why don't I have more control?”

3) Multiresolution Language Design

Our Approach: Structure the language in a layered manner, permitting it to be used at multiple levels as required/desired

- provide high-level features and automation for convenience
- provide the ability to drop down to lower, more manual levels
- use appropriate separation of concerns to keep these layers clean



4) Ability to Tune for Locality/Affinity

- Large-scale systems tend to locate memory w/ processors
 - a good approach for building scalable parallel systems
- Remote accesses tend to be significantly more expensive than local
- Therefore, placement of data relative to computation matters for scalable performance
 - ⇒ programmer should have control over placement of data, tasks
- As multicore chips grow in #cores, locality likely to become more important in mainstream parallel programming as well
 - GPUs/accelerators are another case where locality matters

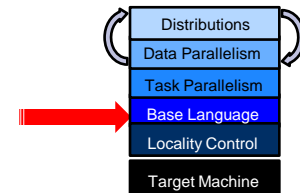
5) Support for Modern Language Concepts

- students graduate with training in Java, Matlab, Perl, C#
- HPC community mired in Fortran, C (maybe C++) and MPI
- we'd like to narrow this gulf
 - leverage advances in modern language design
 - better utilize the skills of the entry-level workforce...
...while not ostracizing traditional HPC programmers
- examples:
 - build on an imperative, block-structured language design
 - support object-oriented programming, but make its use optional
 - support for static type inference, generic programming to support...
...exploratory programming as in scripting languages
...code reuse

Outline

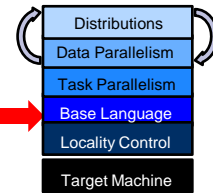
- ✓ Chapel Context
- ✓ Chapel Themes
- Language Overview
 - Base Language
 - Task Parallelism
 - Data Parallelism
 - Locality and Distributions
- Status, Collaborations, Future Work

Base Language: Design



- 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
- Follow lead of C family of languages when useful
(C, Java, C#, Perl, ...)

Base Language: My Favorite Features



■ Rich compile-time language

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

■ Latent types

- ability to omit type specifications for convenience or code 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)

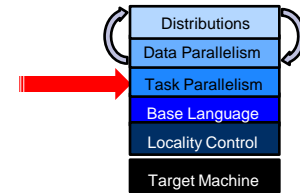
```
config const n = 100; // override with ./a.out --n=100000
```

■ Tuples

■ Iterators (in the CLU, Ruby sense)

■ Declaration Syntax: more like Pascal/Modula than C

Task Parallelism: Task Creation



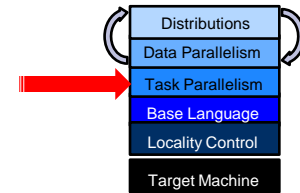
begin: creates a task for future evaluation

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

sync: waits on all begins created within its dynamic scope

```
sync {  
  begin treeSearch(root);  
}  
  
def treeSearch(node) {  
  if node == nil then return;  
  begin treeSearch(node.right);  
  begin treeSearch(node.left);  
}
```

Task Parallelism: Structured Tasks



cobegin: creates a task per component statement:

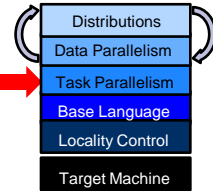
```
computePivot(lo, hi, data);
cobegin {
    Quicksort(lo, pivot, data);
    Quicksort(pivot, hi, data);
} // implicit join here
```

```
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$: sync real; // result is initially empty
sync {
  begin ... = result$; // block until full, leave empty
  begin result$ = ...; // block until empty, leave full
}
result$.readXX(); // read value, leave state unchanged;
// other variations also supported
```

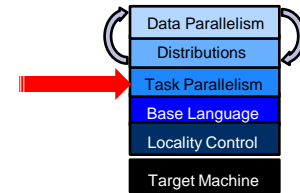
single-assignment variables: writable once only

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

atomic sections: support transactions against memory

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

Producer/Consumer example



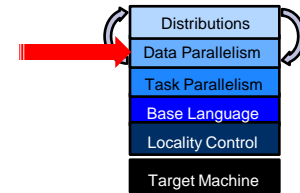
```
var buff$: [0..bufferize-1] sync int;
```

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

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

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

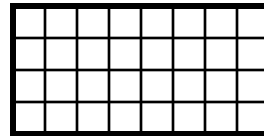
Data Parallelism: Domains



domain: a first-class index set

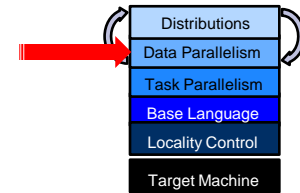
```
var m = 4, n = 8;
```

```
var D: domain(2) = [1..m, 1..n];
```



D

Data Parallelism: Domains

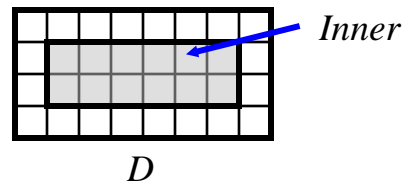


domain: a first-class index set

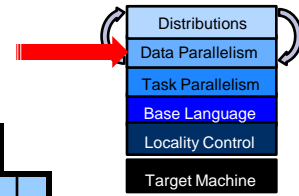
```
var m = 4, n = 8;
```

```
var D: domain(2) = [1..m, 1..n];
```

```
var Inner: subdomain(D) = [2..m-1, 2..n-1];
```

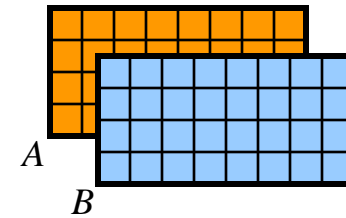


Domains: Some Uses



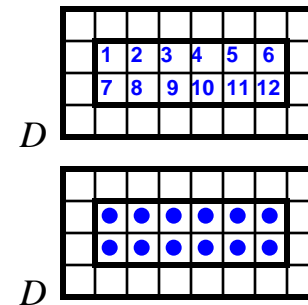
- Declaring arrays:

```
var A, B: [D] real;
```



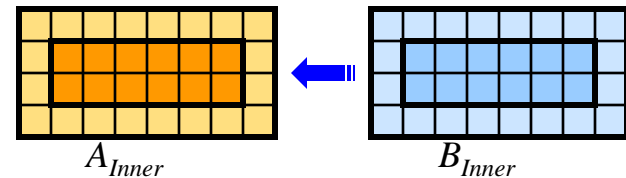
- Iteration (sequential or parallel):

```
for ij in Inner { ... }
or: forall ij in Inner { ... }
or: ...
```



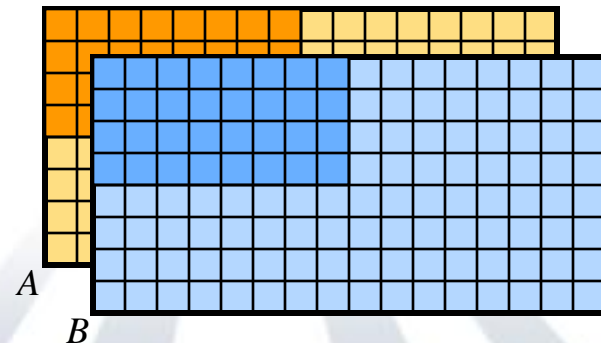
- Array Slicing:

```
A[Inner] = B[Inner];
```

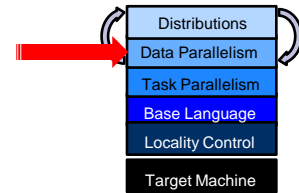


- Array reallocation:

```
D = [1..2*m, 1..2*n];
```

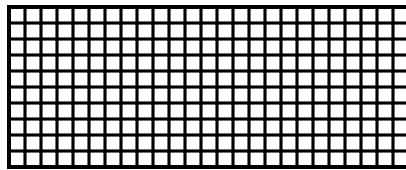


Data Parallelism: Domain Types

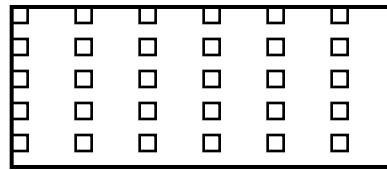


Chapel supports several domain types...

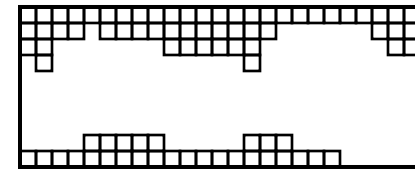
```
var OceanSpace = [0..#lat, 0..#long],
    AirSpace = OceanSpace by (2,4),
    IceSpace: sparse subdomain(OceanSpace) = genCaps();
```



dense

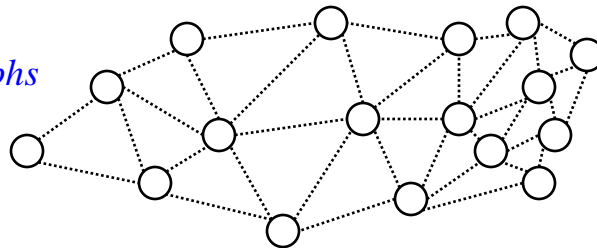


strided

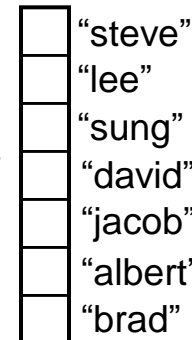


sparse

graphs

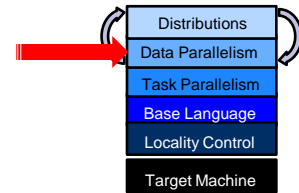


associative



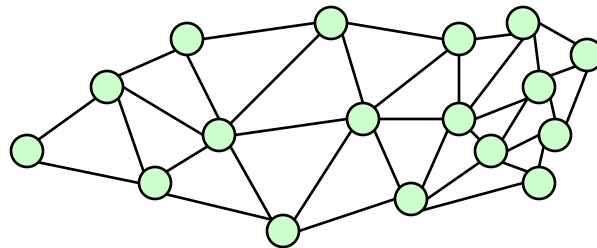
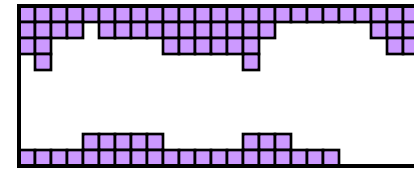
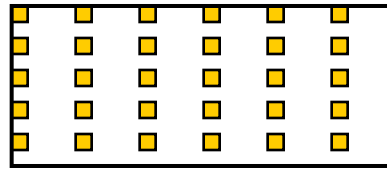
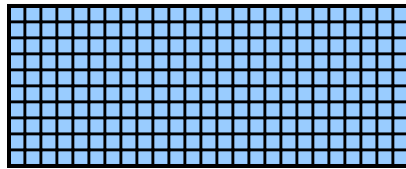
```
var Vertices: domain(opaque) = ..., People: domain(string) = ...;
```

Data Parallelism: Domain Uses



All domain types can be used to declare arrays...

```
var Ocean: [OceanSpace] real,
    Air: [AirSpace] real,
    IceCaps[IceSpace] real;
```

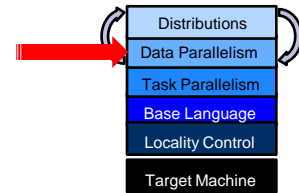


- “steve”
- “lee”
- “sung”
- “david”
- “jacob”
- “albert”
- “brad”

```
var Weight: [Vertices] real,
```

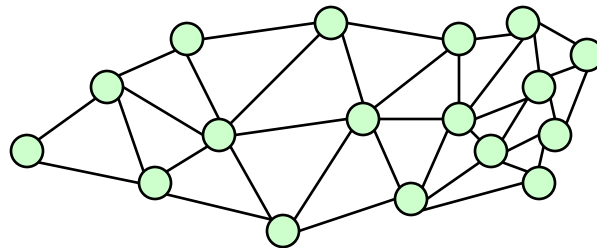
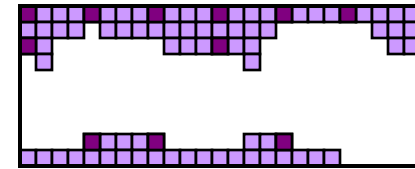
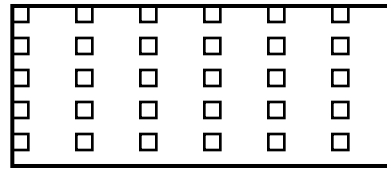
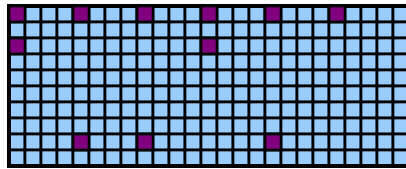
```
Age: [People] int;
```

Data Parallelism: Domain Uses



...to iterate over index sets...

```
forall ij in AirSpace do
    Ocean(ij) += IceCaps(ij);
```

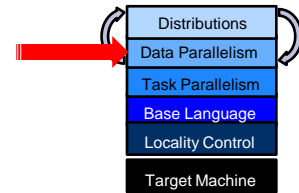


- “steve”
- “lee”
- “sung”
- “david”
- “jacob”
- “albert”
- “brad”

```
forall v in Vertices do
    Weight(v) = numEdges(v);
```

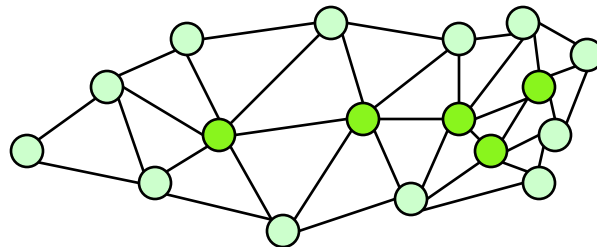
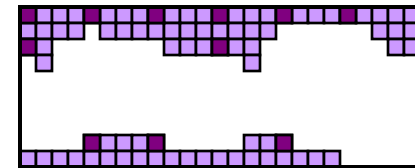
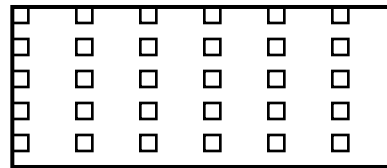
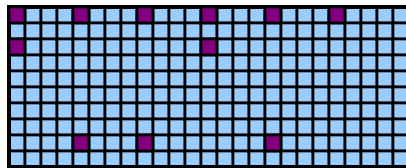
```
forall p in People do
    Age(p) += 1;
```


Data Parallelism: Domain Uses



...to slice arrays...

```
Ocean[AirSpace] += IceCaps[AirSpace];
```

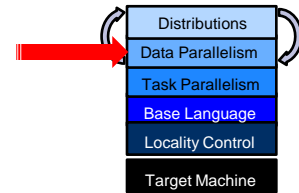


- “steve”
- “lee”
- “sung”
- “david”
- “jacob”
- “albert”
- “brad”

...Vertices[Interior]...

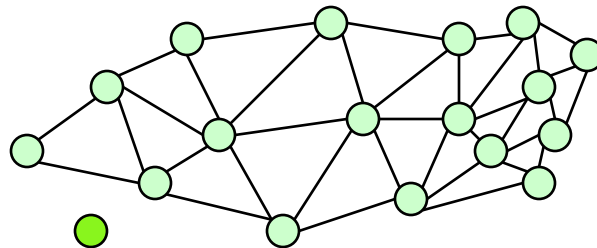
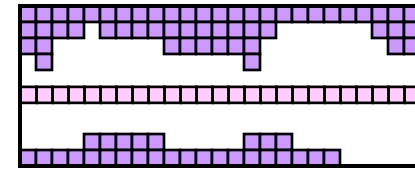
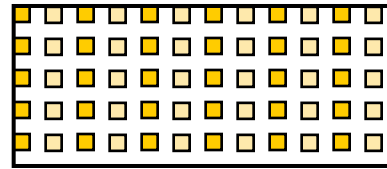
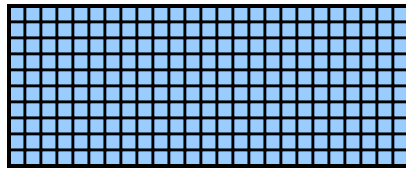
...People[Interns]...

Data Parallelism: Domain Uses



...and to reallocate arrays

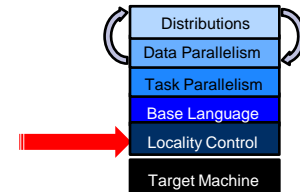
```
AirSpace = OceanSpace by (2,2);
IceSpace += genEquator();
```



- “steve”
- “lee”
- “sung”
- “david”
- “jacob”
- “albert”
- “brad”
- “srini”

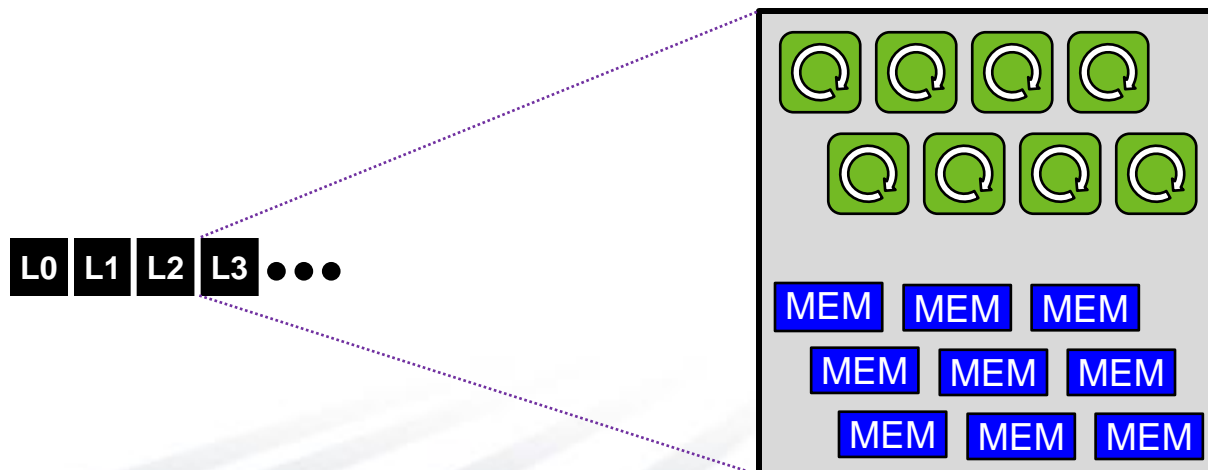
```
newnode = Vertices.create();    People += “srini”;
```

Locality: Locales

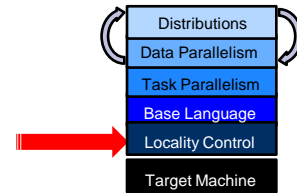


locale: An abstract unit of the target architecture

- supports reasoning about locality
- has capacity for processing and storage
- two threads in a given locale have similar access to a given address
 - addresses in that locale are ~uniformly accessible
 - addresses in other locales are also accessible, but at a price
- locales are defined for a given architecture by a Chapel compiler
 - e.g., a multicore processor or SMP node could be a locale



Locales and Program Startup



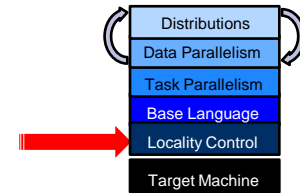
- Chapel users specify # locales on executable command-line

```
prompt> myChapelProg -nl=8           # run using 8 locales
```

L0 L1 L2 L3 L4 L5 L6 L7

- Chapel launcher bootstraps program execution:
 - obtains necessary machine resources
 - e.g., requests 8 nodes from the job scheduler
 - loads a copy of the executable onto the machine resources
 - starts running the program. *Conceptually...*
 - ...locale #0 starts running program's entry point (`main()`)
 - ...other locales wait for work to arrive

Locale Variables

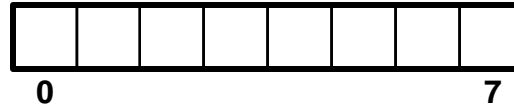


Built-in variables represent a program's locale set:

```
config const numLocales: int;           // number of locales
const LocaleSpace = [0..numLocales-1], // locale indices
      Locales: [LocaleSpace] locale;    // locale values
```

numLocales: 8

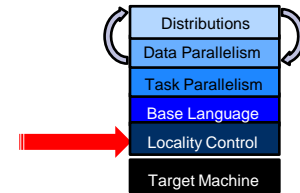
LocaleSpace:



Locales:



Locale Views

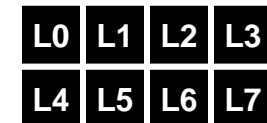


Using standard array operations, users can create their own locale views:

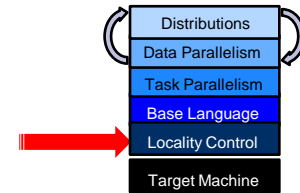
```
var TaskALocs = Locales[..numTaskALocs];
var TaskBLocs = Locales[numTaskALocs+1..];
```



```
var CompGrid = Locales.reshape([1..gridRows,
                                1..gridCols]);
```



Locale Methods



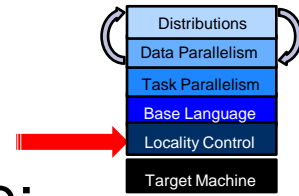
- The locale type supports built-in methods:

```
def locale.id: int;           // index in LocaleSpace
def locale.name: string;     // similar to uname -n
def locale.numCores: int;    // # of processor cores
def locale.physicalMemory(...): ...; // amount of memory
...
```

- Locale queries can also be made:

```
...myvar.locale... // query the locale where myvar is stored
...here...         // query where the current task is running
```

Locality: Task Placement



on clauses: indicate where statements should execute:

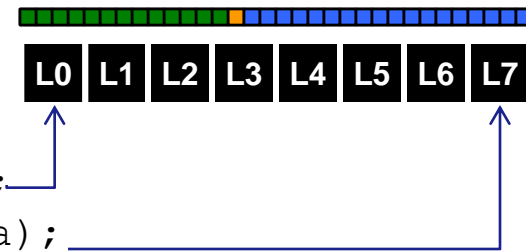
Either by naming locales explicitly...

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



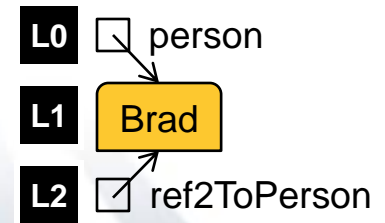
...or in a data-driven manner:

```
const pivot = computePivot(lo, hi, data);
cobegin {
  on data[lo] do Quicksort(lo, pivot, data);
  on data[hi] do Quicksort(pivot+1, hi, data);
}
```

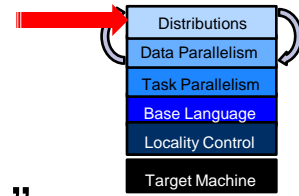


They can also control where data is allocated:

```
var person: Employee;
on Locales(1) do person = new Employee("Brad");
on Locales(2) do var ref2ToPerson = person;
```

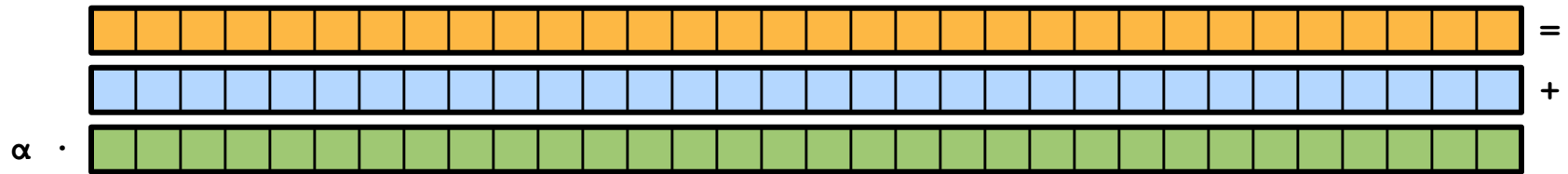


Chapel Distributions

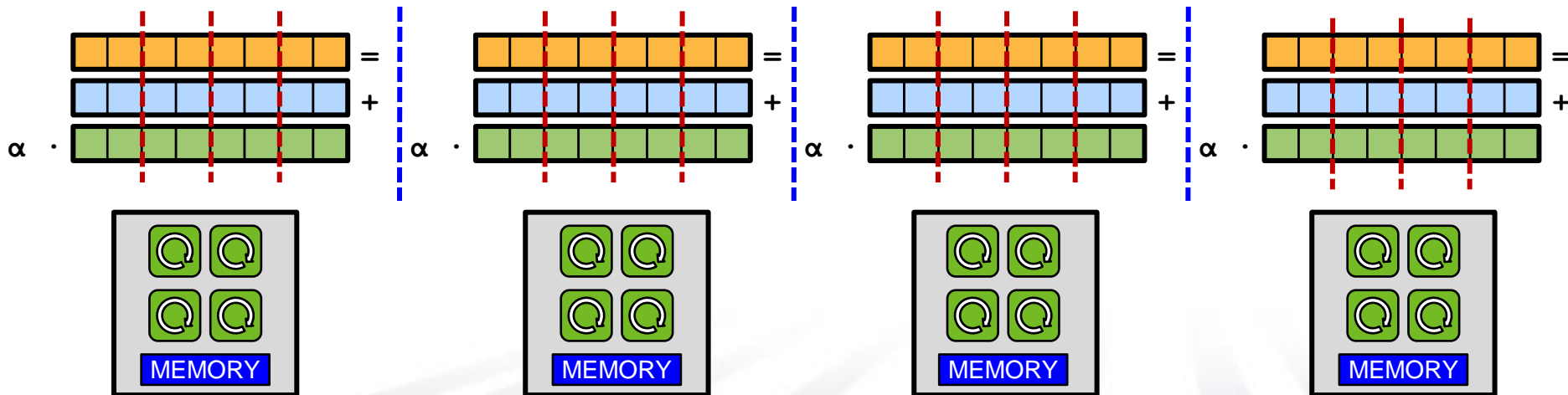


Distributions: “Recipes for parallel, distributed arrays”

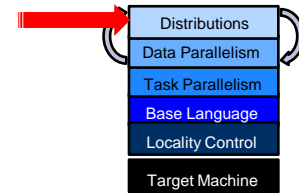
- help the compiler map from the computation’s global view...



...down to the *fragmented*, per-processor implementation

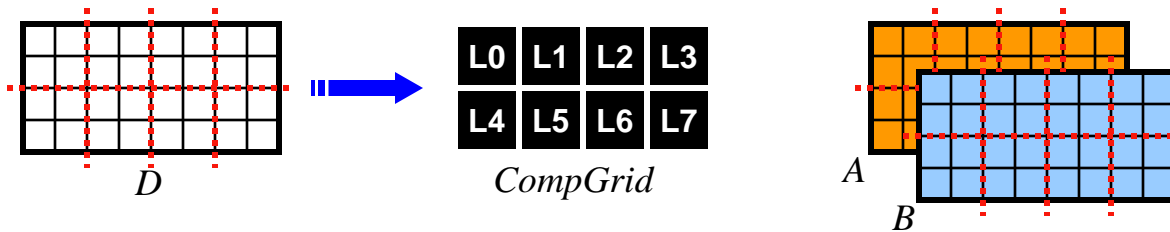


Domain Distribution



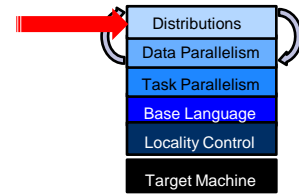
Domains may be distributed across locales

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



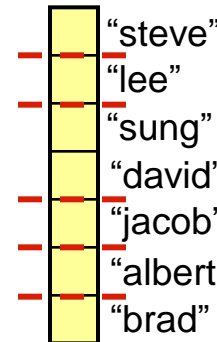
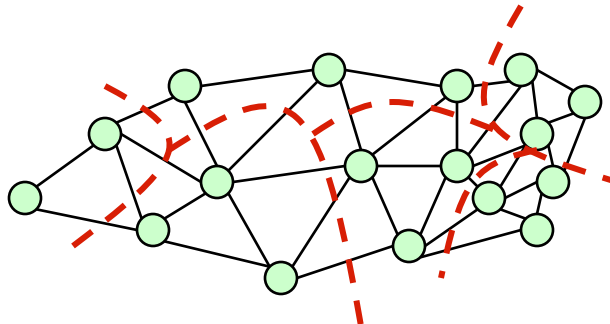
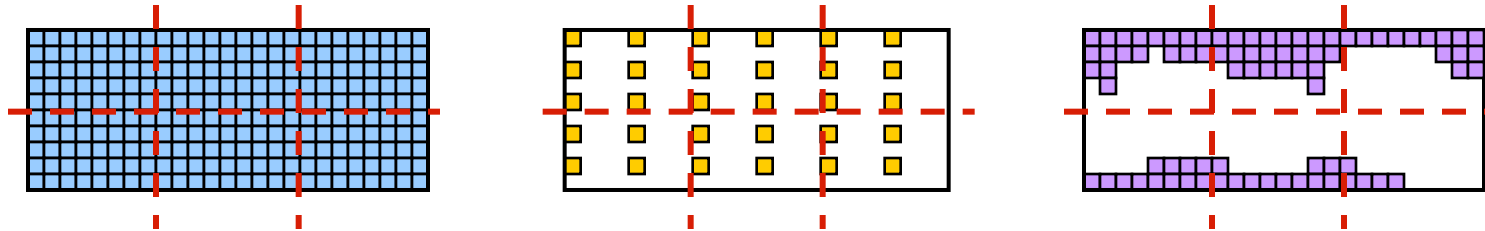
A distribution implies...

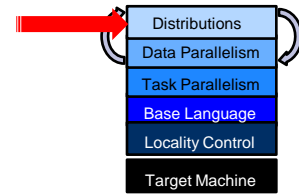
- ...ownership of the domain's indices (and its arrays' elements)
- ...the default work ownership for operations on the domains/arrays
 - e.g., forall loops or promoted operations
- ...the implementation of operations on its domains and arrays



Domain Distributions

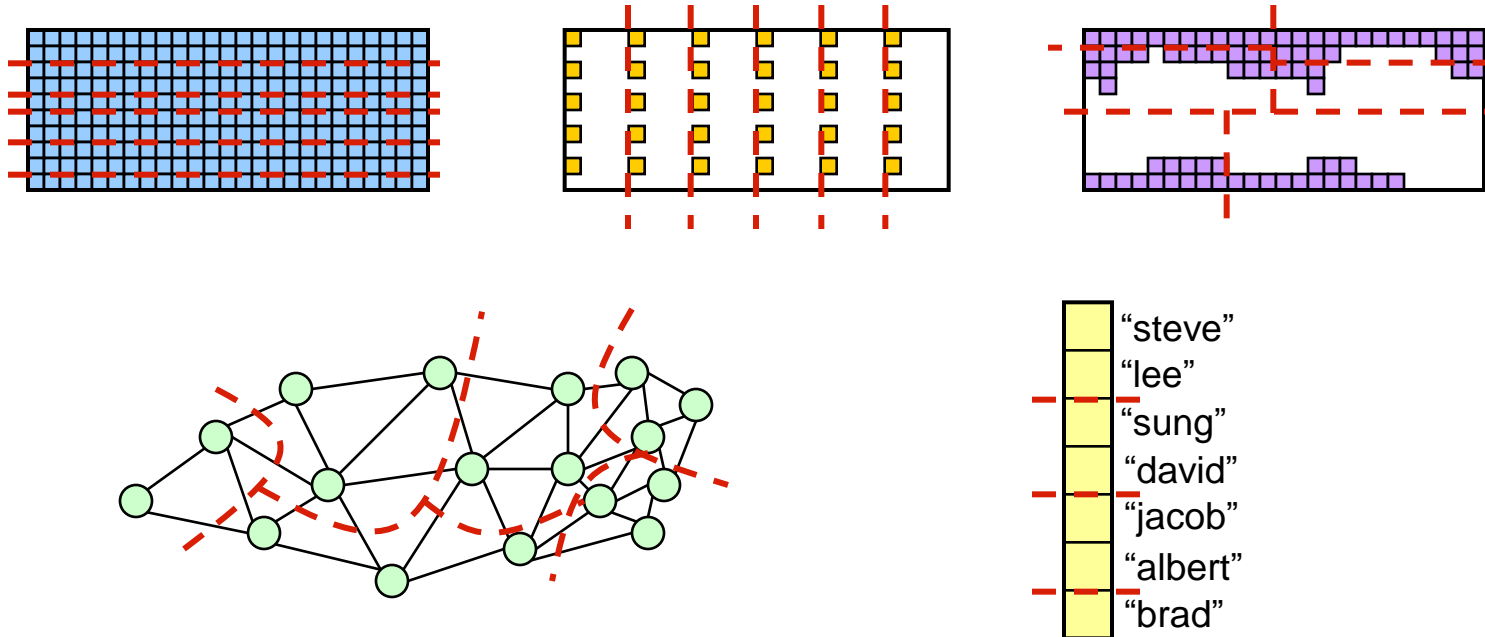
- Any domain type may be distributed
- Distributions do not affect program semantics
 - only implementation details and therefore performance





Domain Distributions

- Any domain type may be distributed
- Distributions do not affect program semantics
 - only implementation details and therefore performance



Distributions: Goals & Research

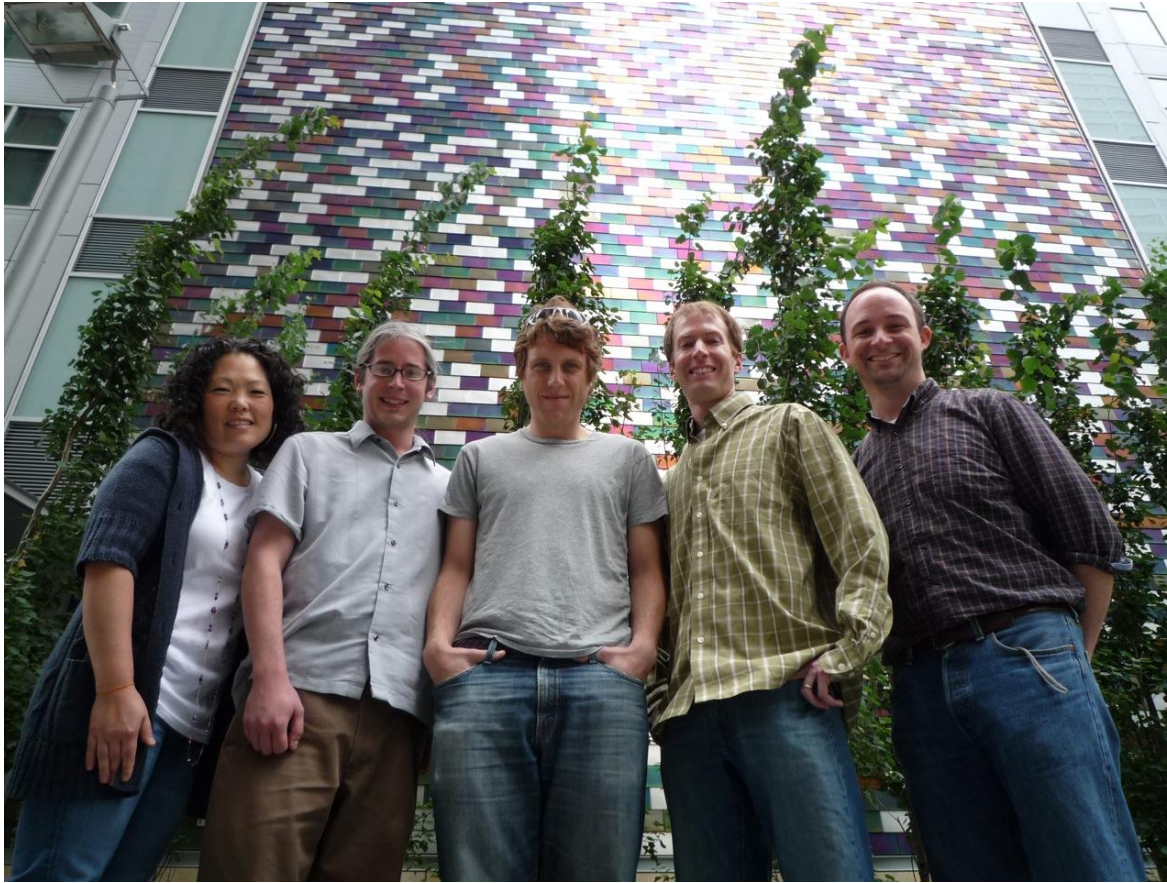
- Advanced users can write their own distributions
 - specified in Chapel using lower-level language features
- Chapel will provide a standard library of distributions
 - written using the same user-defined distribution mechanism

(Draft paper describing user-defined distribution strategy available by request)

Outline

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

The Chapel Team



■ Interns

- Jacob Nelson ('09 – UW)
- Albert Sidelnik ('09 – UIUC)
- Andy Stone ('08 – Colorado St)
- James Dinan ('07 – Ohio State)
- Robert Bocchino ('06 – UIUC)
- Mackale Joyner ('05 – Rice)

■ Alumni

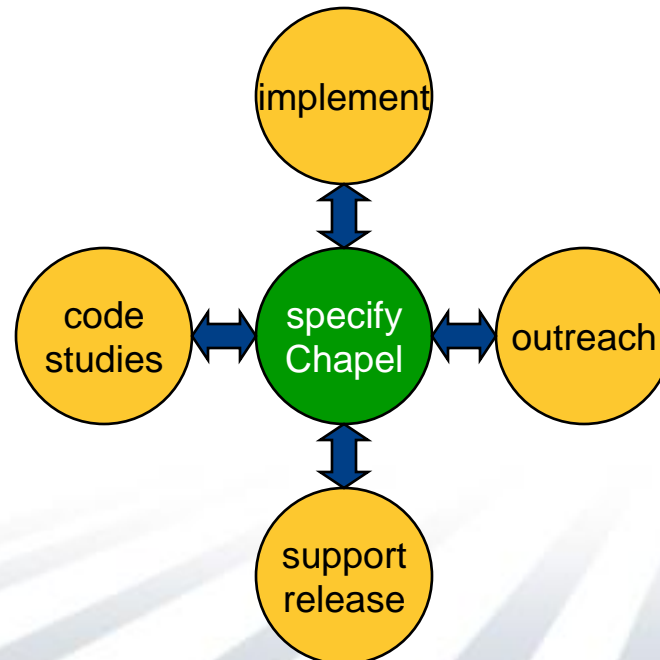
- David Callahan
- Roxana Diaconescu
- Samuel Figueroa
- Shannon Hoffswell
- Mary Beth Hribar
- Mark James
- John Plevyak
- Wayne Wong
- Hans Zima

Sung-Eun Choi, David Iten, Lee Prokowich,
Steve Deitz, Brad Chamberlain

Chapel Work

■ Chapel Team's Focus:

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



Chapel Release

- **Current release:** version 1.0 (October 15th, 2009)
- Supported environments: UNIX/Linux, Mac OS X, Cygwin
- How to get started:
 1. Download from: <http://sourceforge.net/projects/chapel>
 2. Unpack tar.gz file
 3. See top-level README
 - for quick-start instructions
 - for pointers to next steps with the release
- Your feedback desired!
- **Remember:** a work-in-progress
 - ⇒ it's likely that you will find problems with the implementation
 - ⇒ this is still a good time to influence the language's design

Implementation Status (v1.0)

- **Base language:** stable (some gaps and bugs remain)
- **Task parallel:**
 - stable multi-threaded implementation of tasks, sync variables
 - atomic sections are an area of ongoing research with U. Notre Dame
- **Data parallel:**
 - stable multi-threaded data parallelism for dense domains/arrays
 - other domain types have a single-threaded reference implementation
- **Locality:**
 - stable locale types and arrays
 - stable task parallelism across multiple locales
 - initial support for some distributions: Block, Cyclic, Block-Cyclic
- **Performance:**
 - has received much attention in designing the language
 - yet not much implementation effort to date

Outreach: Active Collaborations

Notre Dame/ORNL (Peter Kogge, Srinivas Sridharan, Jeff Vetter):

Asynchronous STM over distributed memory

UIUC (David Padua, Albert Sidelnik):

Chapel for hybrid CPU-GPU computing

OSU (Gagan Agrawal, Bin Ren):

Data-intensive computing using Chapel's user-defined reductions

PNNL/CASS-MT (John Feo, Daniel Chavarria): Chapel extensions for hybrid computation; performance tuning for the Cray XMT; ARMCI port

ORNL (David Bernholdt *et al.*; Steve Poole *et al.*): Chapel code studies – Fock matrix computations, MADNESS, Sweep3D, coupled models, ...

Berkeley (Dan Bonachea *et al.*): APGAS over GASNet; collectives

(Your name here?)

Collaboration Opportunities

- memory management policies/mechanisms
- dynamic load balancing: task throttling and stealing
- parallel I/O and checkpointing
- language interoperability
- application studies and performance optimizations
- index/subdomain semantics and optimizations
- targeting different back-ends (LLVM, MS CLR, ...)
- runtime compilation
- library support
- tools
 - correctness debugging
 - performance debugging
 - IDE support
 - Chapel interpreter
 - visualizations, algorithm animations
- (your ideas here...)

Next Steps

- Expand our set of supported distributions
- Continue to improve performance
- Continue to add missing features
- Expand the set of codes that we are studying
- Expand the set of architectures that we are targeting
- Support the public release
- Continue to support collaborations and seek out new ones

Summary

Chapel strives to greatly improve Parallel Productivity

via its support for...

- ...general parallel programming
- ...global-view abstractions
- ...control over locality
- ...multiresolution features
- ...modern language concepts and themes

For More Information

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<http://chapel.cray.com>

<http://sourceforge.net/projects/chapel>

Parallel Programmability and the Chapel Language;
Chamberlain, Callahan, Zima; International Journal of High
Performance Computing Applications, August 2007,
21(3):291-312.

Questions?

