Chapelthe 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

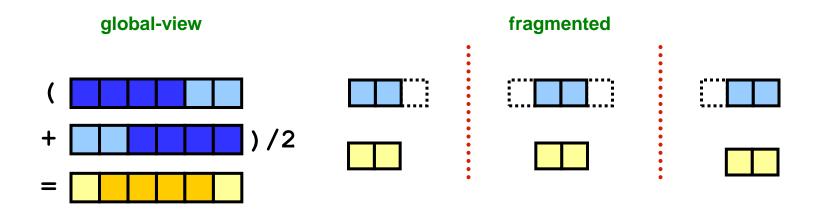
- Types: multicore desktops, clusters, HPC systems, ...
- Levels: inter-machine, inter-node, inter-core, vectors, multithreading





2) Global-view vs. Fragmented

Problem: "Apply 3-pt stencil to vector"

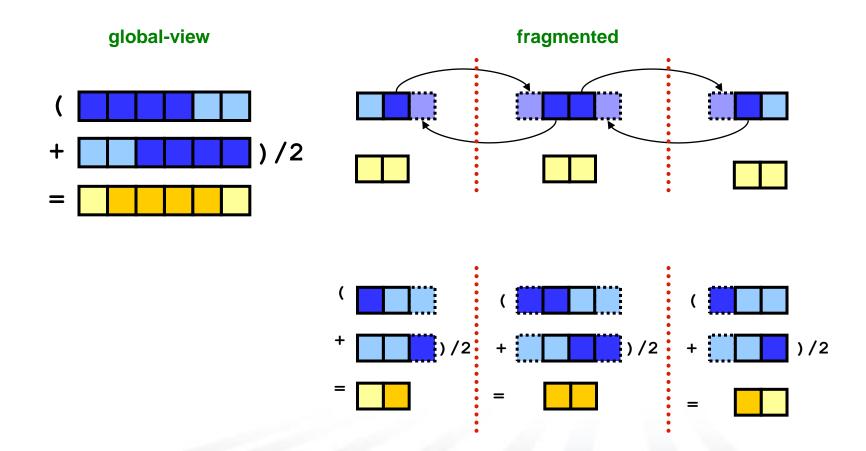






2) Global-view vs. Fragmented

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));
    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;
```





2) SPMD pseudo-code + MPI

Problem: "Apply 3-pt stencil to vector"

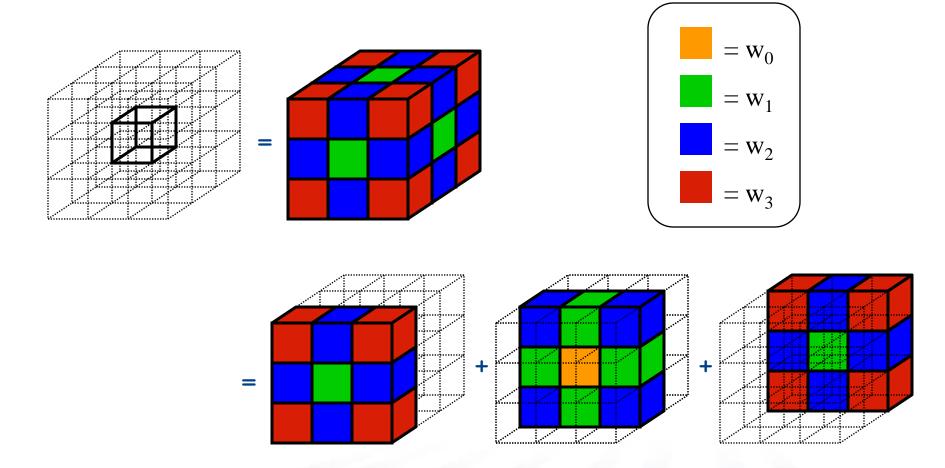
SPMD (pseudocode + MPI)

```
var n: int = 1000, locN: int = n/numProcs;
var a, b: [0..locN+1] real;
                                                               Communication becomes
var innerLo: int = 1, innerHi: int = locN;
                                                            geometrically more complex for
var numProcs, myPE: int;
                                                               higher-dimensional arrays
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) {</pre>
  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;
```

ARPA HPES



2) rprj3 stencil from NAS MG







2) NAS MG rprj3 stencil in Fortran + MPI

```
subroutine comm3(u,n1,n2,n3,kk)
use caf intrinsics
implicit none
include 'cafnpb.h'
integer n1, n2, n3, kk
integer axis
if( .not. dead(kk) )ther
   do axis = 1, 3
         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)
         call commlp(axis, u, n1, n2, n3, kk)
      endif
      call sync_all()
      call sync_all()
   call zero3(u.n1.n2.n3)
return
subroutine give3 ( axis, dir, u, n1, n2, n3, k )
implicit none
include 'cafnpb.h
integer axis, dir, n1, n2, n3, k, ierr
integer i3, i2, i1, buff len,buff id
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)
      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
            buff len = buff len + 1
            buff(buff_len, buff_id) = u(n1-1, i2,i3)
      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
            buff len = buff len + 1
            buff(buff_len, buff_id) = u(i1, 2,i3)
 buff(1:buff len,buff id+1)[nbr(axis,dir,k)] =
```

```
else if( dir .eg. +1 ) then
         do i1=1.n1
            buff len = buff len + 1
            buff(buff len, buff id )= u( i1,n2-1,i3)
      buff(1:buff_len,buff_id+1)[nbr(axis,dir,k)] =
     buff(1:buff len,buff id)
if( axis .eq. 3 ) then if( dir .eq. -1 ) then
      do i2=1.n2
            buff len = buff len + 1
            buff(buff_len, buff_id) = u(i1,i2,2)
      buff(1:buff len.buff id+1)[nbr(axis.dir.k)] =
      buff(1:buff len,buff id)
   else if (dir .eq. +1 ) ther
      do i2=1.n2
         do i1=1.n1
            buff_len = buff_len + 1
            buff(buff len, buff id) = u(i1,i2,n3-1)
       buff(1:buff len,buff id+1)[nbr(axis,dir,k)] =
      buff(1:buff len,buff id)
   endif
return
subroutine take3 (axis, dir, u, n1, n2, n3)
use caf intrinsics
implicit none
include 'cafnpb.h'
integer axis, dir, n1, n2, n3
double precision u( n1, n2, n3 )
integer buff id, indx
integer i3, i2, i1
buff id = 3 + dir
if( axis .eq. 1 ) then
if( dir .eq. -1 ) then
         do i2=2.n2-1
            u(n1,i2,i3) = buff(indx, buff id )
   else if( dir .eq. +1 ) then
      do i3=2.n3-1
            indx = indx + 1
            u(1,i2,i3) = buff(indx, buff id)
         enddo
   endif
if(axis.eq. 2)then
   if ( dir .eq. -1 ) then
      do i3=2,n3-1
do i1=1,n1
            u(i1,n2,i3) = buff(indx, buff id)
```

```
else if( dir .eq. +1 ) then
      do i3=2,n3-1
         do i1=1.n1
            u(i1,1,i3) = buff(indx, buff id)
endif
if( axis .eq. 3 )then
   if( dir .eq. -1 )then
      do i2=1.n2
            u(i1.i2.n3) = buff(indx, buff id)
   else if( dir .eq. +1 ) then
            u(i1,i2,1) = buff(indx, buff_id)
   endif
endif
subroutine commlp(axis, u, n1, n2, n3, kk)
include 'globals h
integer axis, dir, n1, n2, n3
double precision u( n1, n2, n3 )
integer i3, i2, i1, buff len, buff id
buff id = 3 + dir
do i=1.nm2
  buff(i,buff_id) = 0.0D0
enddo
dir = +1
buff id = 3 + dir
do i=1 nm2
   buff(i,buff_id) = 0.0D0
enddo
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,
   12.131
   enddo
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-
      enddo
```

```
if( axis .eg. 3 )then
   do i2=1,n2
do i1=1,n1
         buff_len = buff_len + 1
buff(buff len, buff id) = u(i1,i2,n3-
      enddo
buff id = 2 + dir
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(2, i2,i3)
       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,
    2,13)
      enddo
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,i2,2)
endif
do i=1.nm2
   buff(i,4) = buff(i,3)
buff(i,2) = buff(i,1)
buff id = 3 + dir
if( axis .eq. 1 ) then
do i3=2,n3-1
      do i2=2,n2-1
          u(n1,i2,i3) = buff(indx, buff id)
       enddo
if( axis .eq. 2 )then
   do i3=2,n3-1
       do i1=1,n1
          indx = indx + 1
          u(i1,n2,i3) = buff(indx, buff id)
    enddo
if( axis .eq. 3 )then do i2=1,n2
      do i1=1,n1
          u(i1,i2,n3) = buff(indx, buff id)
       enddo
endif
buff id = 3 + dir
if( axis .eq. 1 )then
do i3=2,n3-1
      do i2=2,n2-1
          u(1,i2,i3) = buff(indx, buff id)
```

endif

```
if (axis .eq. 2 ) then
  do i3=2,n3-1
     do i1=1,n1
        indx = indx + 1
         u(i1,1,i3) = buff(indx, buff id)
     enddo
endif
if(axis.eq. 3)then
  do i2=1,n2
     do i1=1,n1
        indx = indx + 1
        u(i1,i2,1) = buff(indx, buff_id)
     enddo
  enddo
return
subroutine rpri3(r.mlk.m2k.m3k.s.mli.m2i.m3i.k)
implicit none
include 'globals.h'
integer m1k, 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
 d1 = 2
else
 d1 = 1
endif
 d2 = 2
else
 42 = 1
endif
if (m3k.eq.3) then
 d3 = 2
else
 d3 = 1
andi f
do j3=2,m3j-1
 i3 = 2*i3-d3
 do j2=2,m2j-1
    do 11=2.m11
      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)
    do j1=2,m1j-1
     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(11,12,13) =
          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
  4 = k-1
  call comm3(s,m1j,m2j,m3j,j)
```





2) NAS MG rprj3 stencil in Chapel

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



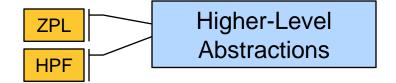


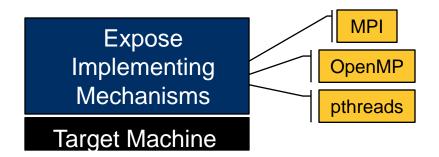


3) Multiresolution Languages: Motivation

Two typical camps of parallel language design:

low-level vs. high-level





Target Machine

"Why is everything so tedious?"

"Why don't I have more control?"

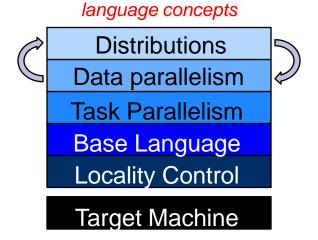




3) Multiresolution Language Design

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



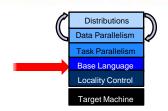


Outline

- √ 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)

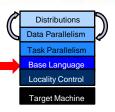
C, Modula, Ada: syntax

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
```

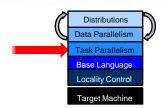
- Tuples
- Iterators (in the CLU, Ruby sense)
- Declaration Syntax: more like Pascal/Modula than C







Task Parallelism: Task Creation



begin: creates a task for future evaluation

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

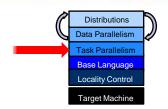
sync: waits on all begins created within its dynamic scope

```
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:

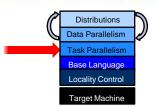
```
computePivot(lo, hi, data);
                                          cobegin {
cobegin {
                                            computeTaskA(...);
  Quicksort(lo, pivot, data);
                                            computeTaskB(...);
  Quicksort(pivot, hi, data);
                    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

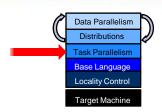
single-assignment variables: writable once only

atomic sections: support transactions against memory

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

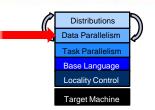


Producer/Consumer example

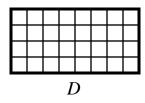


```
var buff$: [0..buffersize-1] sync int;
cobegin {
  producer();
  consumer();
def producer() {
  var i = 0;
  for ... {
    i = (i+1) % buffersize;
    buff$(i) = ...;
def consumer() {
  var i = 0;
  while {
    i = (i+1) % buffersize;
    ...buff$(i)...;
```



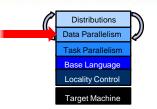


domain: a first-class index set



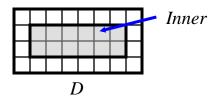






domain: a first-class index set

```
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):

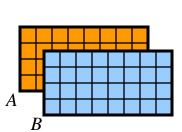
```
ij in Inner { ... }
   for
Or: forall ij in Inner { ... }
or:
```

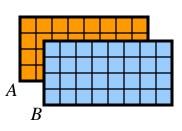


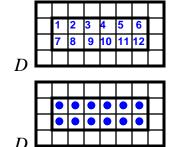
```
A[Inner] = B[Inner];
```

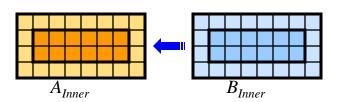
Array reallocation:

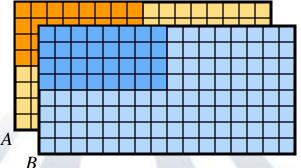
$$D = [1..2*m, 1..2*n];$$





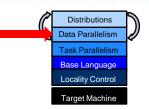












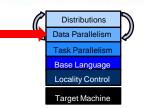
Chapel supports several domain types...

```
var OceanSpace = [0..#lat, 0..#long],
     AirSpace = OceanSpace by (2,4),
     IceSpace: sparse subdomain(OceanSpace) = genCaps();
                                  strided
          dense
                                                         sparse
                                                     "steve"
                                                     "lee"
                                                     "sung"
    graphs
                                         associative
                                                     "david"
                                                     "jacob"
                                                     "albert"
                                                     "brad"
```

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

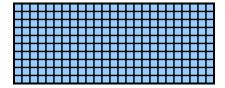


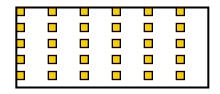


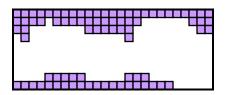


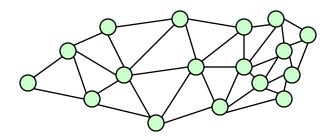
All domain types can be used to declare arrays...

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









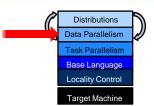
var Weight: [Vertices] real,



Age: [People] int;

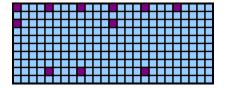


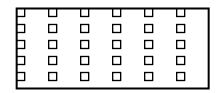


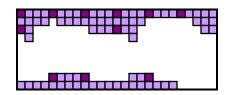


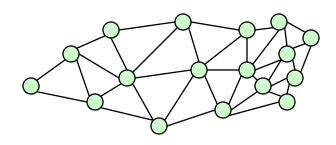
...to iterate over index sets...

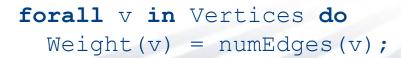
```
forall ij in AirSpace do
Ocean(ij) += IceCaps(ij);
```







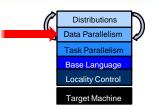






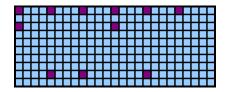


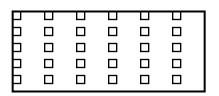


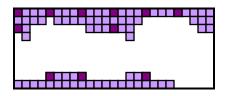


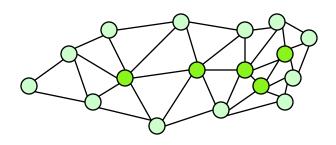
...to slice arrays...

Ocean[AirSpace] += IceCaps[AirSpace];











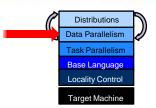
...Vertices[Interior]...

...People[Interns]...



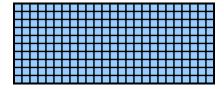


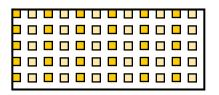


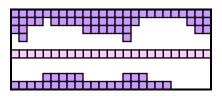


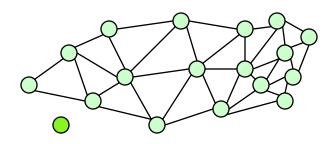
...and to reallocate arrays

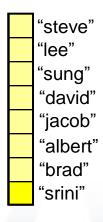
```
AirSpace = OceanSpace by (2,2);
IceSpace += genEquator();
```











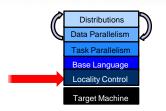
newnode = Vertices.create();

People += "srini";



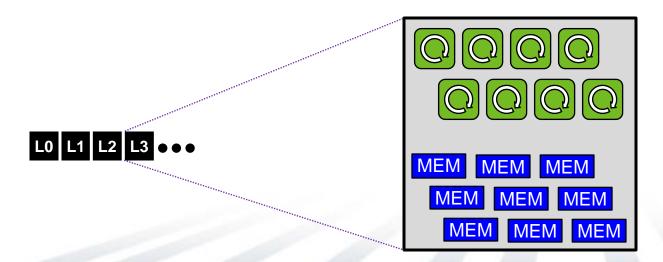


Locality: Locales



locale: An abstract unit of the target architecture

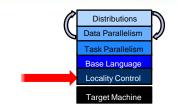
- supports reasoning about locality
- has capacity for processing and storage
- two threads in a given locale have similar access to a given address
 - addresses in that locale are ~uniformly accessible
 - addresses in other locales are also accessible, but at a price
- locales are defined for a given 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

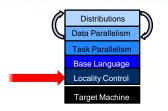
```
prompt> myChapelProg -nl=8  # run using 8 locales
```



- 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
```

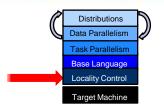
numLocales: 8

LocaleSpace:

Locales: L0 L1 L2 L3 L4 L5 L6 L7



Locale Views

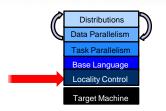


Using standard array operations, users can create their own locale views:

```
L0 L1
var TaskALocs = Locales[..numTaskALocs];
                                                 L2 L3 L4 L5 L6 L7
var TaskBLocs = Locales[numTaskALocs+1..];
var CompGrid = Locales.reshape([1..gridRows,
                                 1..gridCols]);
```



Locale Methods



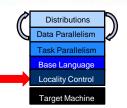
The locale type supports built-in methods:

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



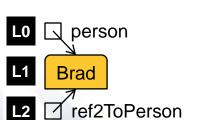
on clauses: indicate where statements should execute:

```
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;
```

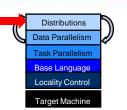






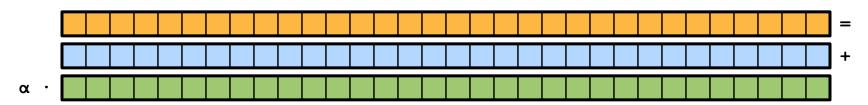


Chapel Distributions

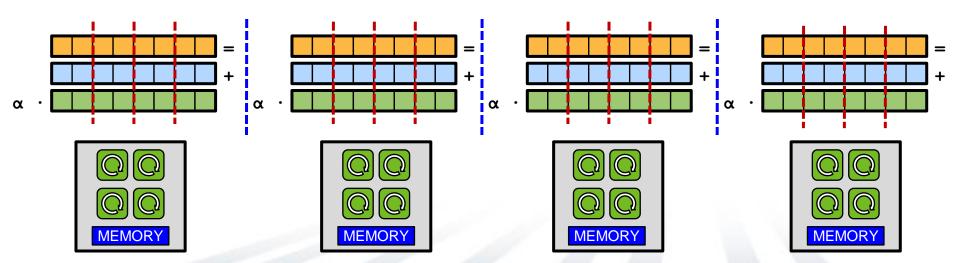


Distributions: "Recipes for parallel, distributed arrays"

help the compiler map from the computation's global view...



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

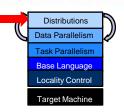






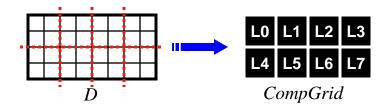


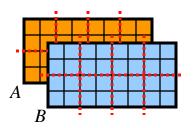
Domain Distribution



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

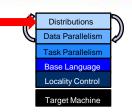
e.g., forall loops or promoted operations

...the implementation of operations on its domains and arrays

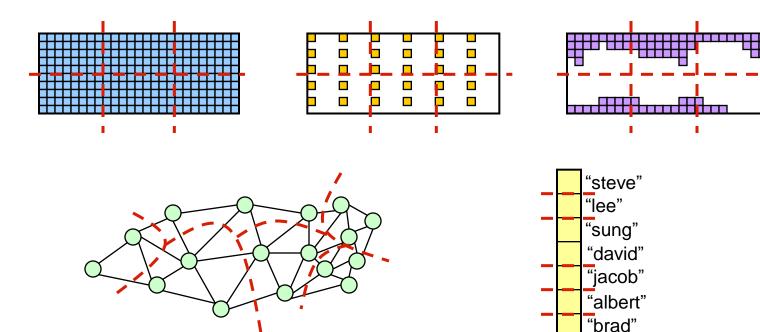




Domain Distributions

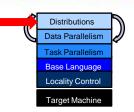


- Any domain type may be distributed
- Distributions do not affect program semantics
 - only implementation details and therefore performance

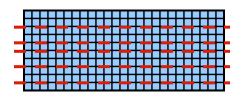


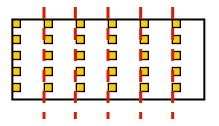


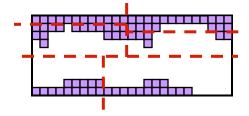
Domain Distributions

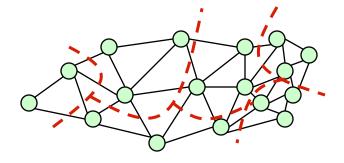


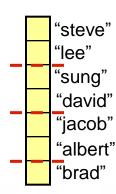
- Any domain type may be distributed
- Distributions do not affect program semantics
 - only implementation details and therefore performance













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)



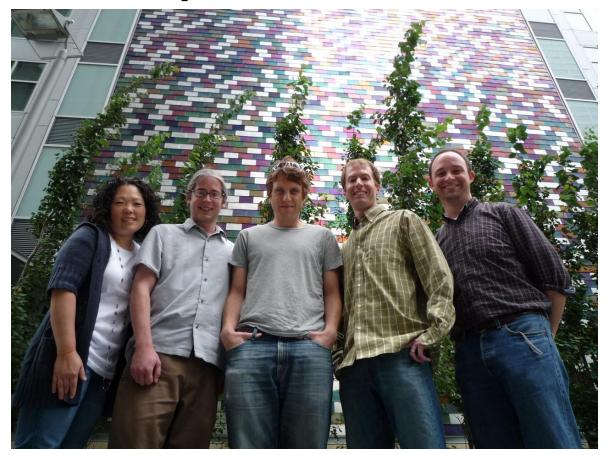


Outline

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



The Chapel Team



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

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

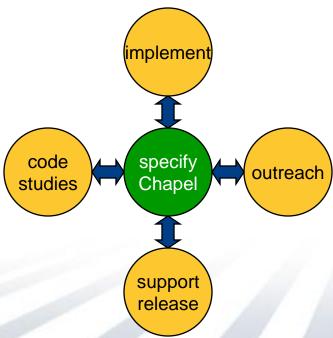
- David Callahan
- Roxana Diaconescu
- Samuel Figueroa
- Shannon Hoffswell
- Mary Beth Hribar
- Mark James
- John Plevyak
- Wayne Wong
- Hans Zima





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...)
 Chapel (54)





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

chapel_info@cray.com

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?





