

Chapel

Productive Parallel Programming at Scale

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Scalability at Cray

Cray's business: Allowing customers to achieve sustained high performance at large machine scales

My job: Enabling users to program our machines more *productively* through innovative language design

Productivity = Performance

- + Programmability
- + Portability
- + Robustness

Scalability Limiters in HPC Programming Models

- Restricted programming and execution models (e.g., SPMD)
- Exposure of low-level implementation mechanisms
- Lack of programmability: ability to _____ code
 - ...write...
 - ...read...
 - ...modify...
 - ...tune...
 - ...maintain...
 - ...experiment with...

Chapel

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

Themes:

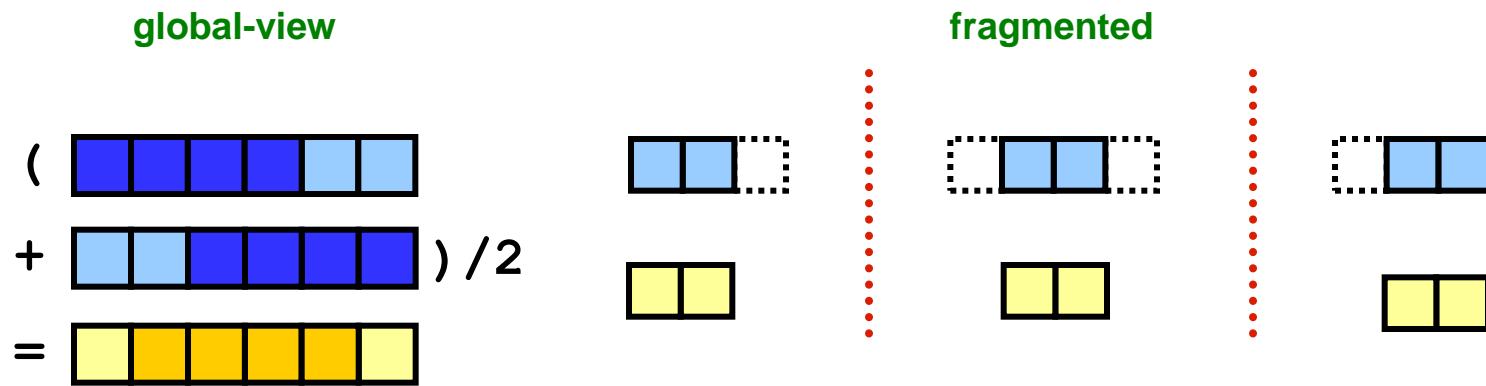
- **general parallel programming**
 - data-, task-, and nested parallelism
 - express general levels of software parallelism
 - target general levels of hardware parallelism
- **reduce gap between mainstream & parallel languages**
- ***global-view abstractions***
- **control of locality**
- ***multiresolution design***

Outline

- ✓ Motivation for Chapel
- Global-view Programming Models and Scalability
- ❑ Language Overview
- ❑ Wrap-up

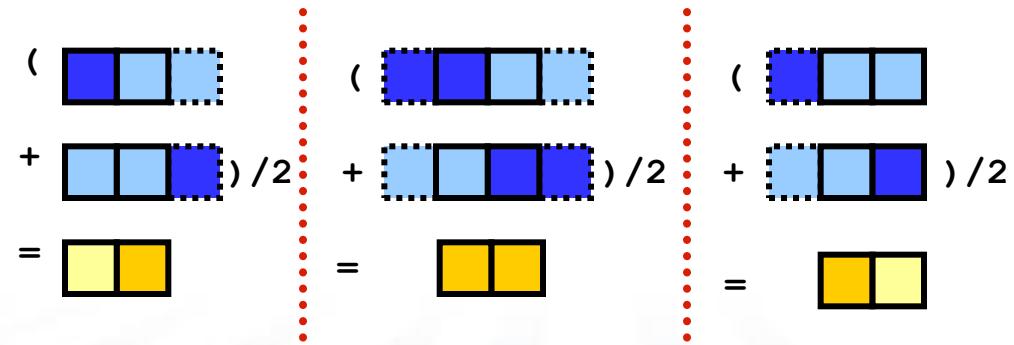
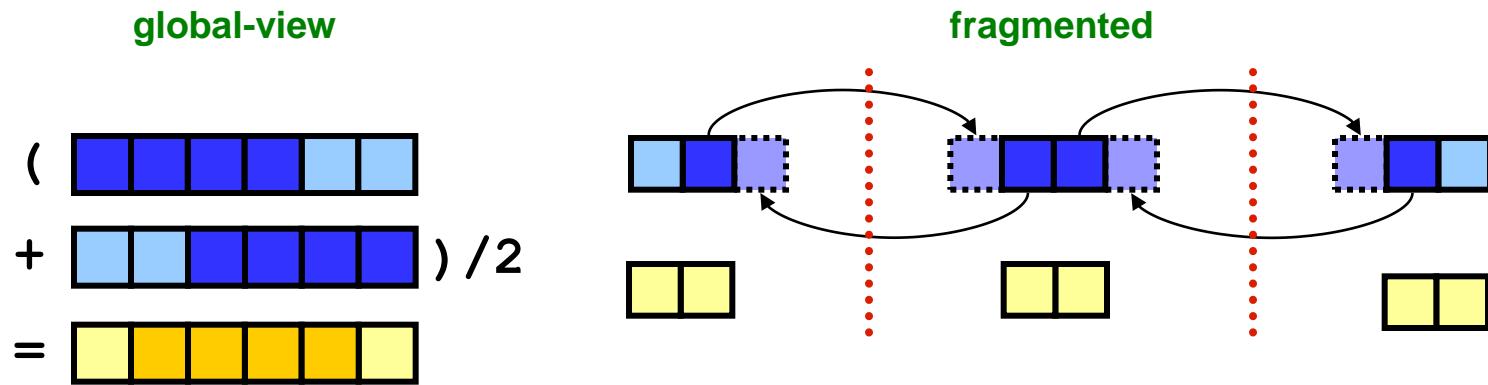
Global-view vs. Fragmented

Problem: “Apply 3-pt stencil to vector”



Global-view vs. Fragmented

Problem: “Apply 3-pt stencil to vector”



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;
    }
}
```

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;
    }
}
```

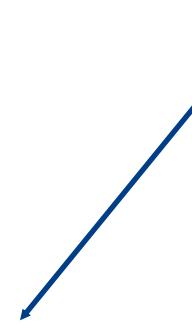
MPI SPMD pseudo-code

Problem: “Apply 3-pt stencil to vector”

SPMD (pseudocode + MPI)

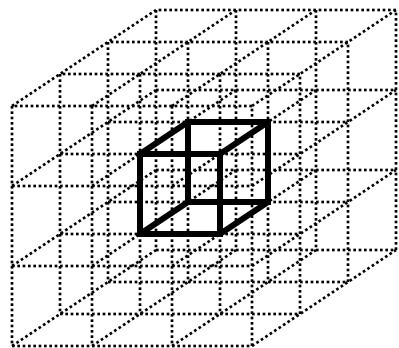
```
var n: int = 1000, locN: int = n/numProcs;
var a, b: [0..locN+1] real;
var innerLo: int = 1, innerHi: int = locN;
var numProcs, myPE: int;
var retval: int;
var status: MPI_Status;

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;
}
```

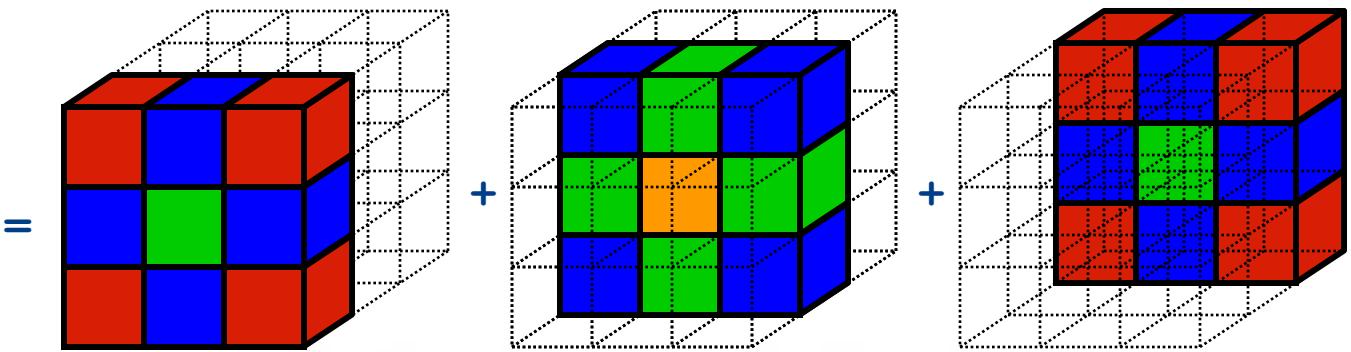
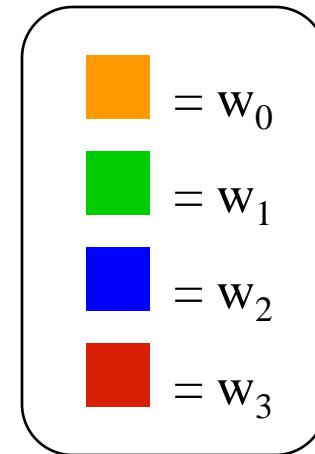
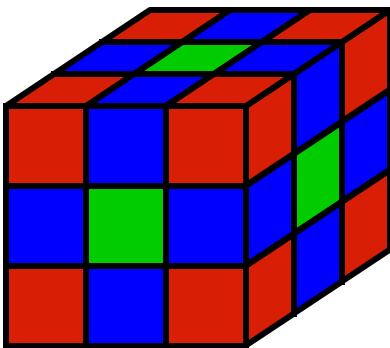


Communication becomes
geometrically more complex for
higher-dimensional arrays

rprj3 stencil from NAS MG



=



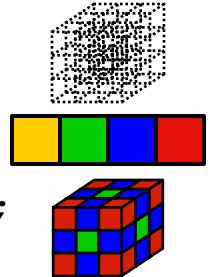
NAS MG *rprj3* stencil in Fortran + MPI

NAS MG *rprj3* stencil in Chapel

```

def rprj3(S, R) {
    param 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 + R.stride*offset));
}

```

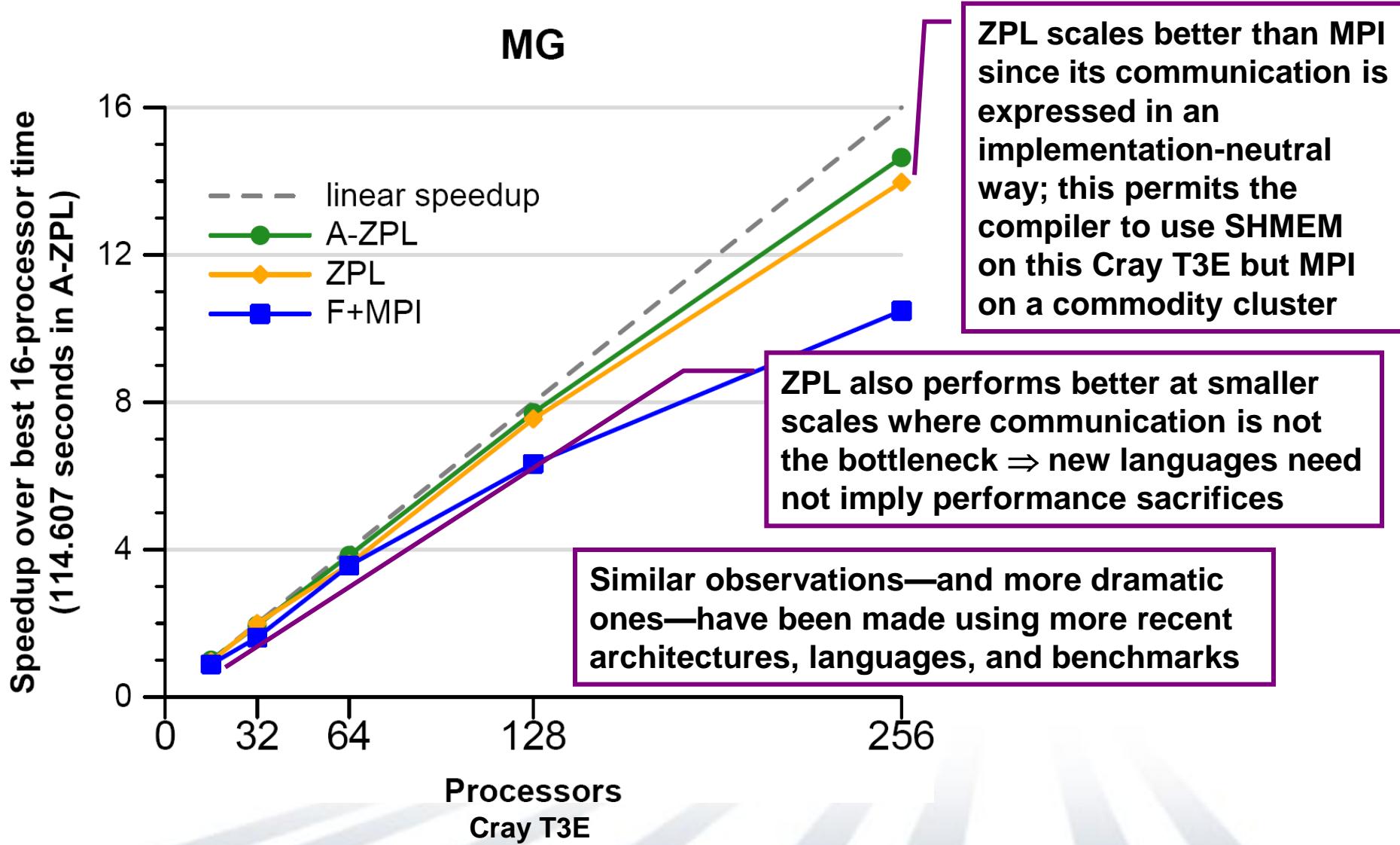


- Unfortunately, Chapel is a work in progress, so I do not have scalability results to show today
- However, these Chapel features are based on our previous work in ZPL, for which I do have scalability results

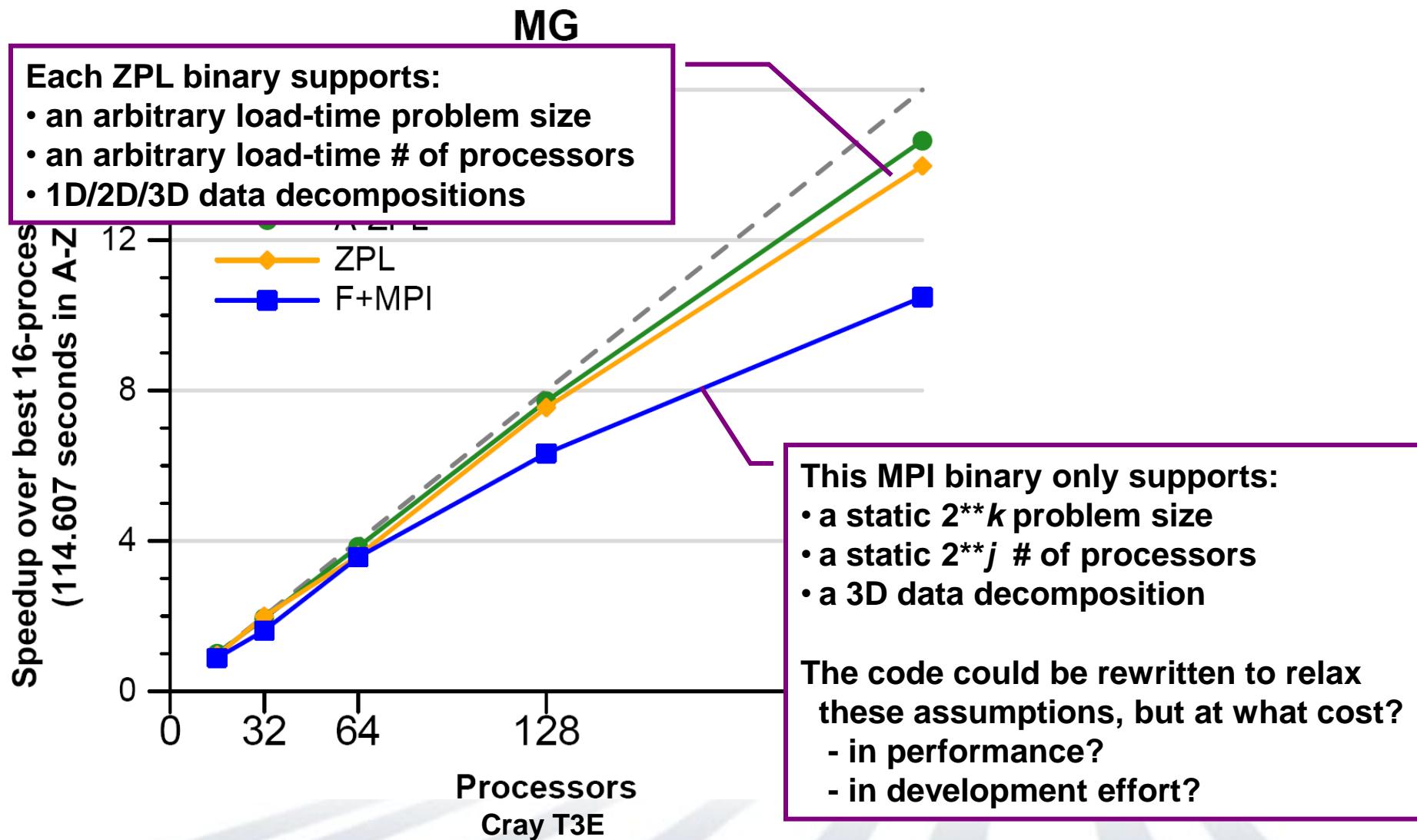
NAS MG *rprj3* stencil in ZPL

```
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;
```

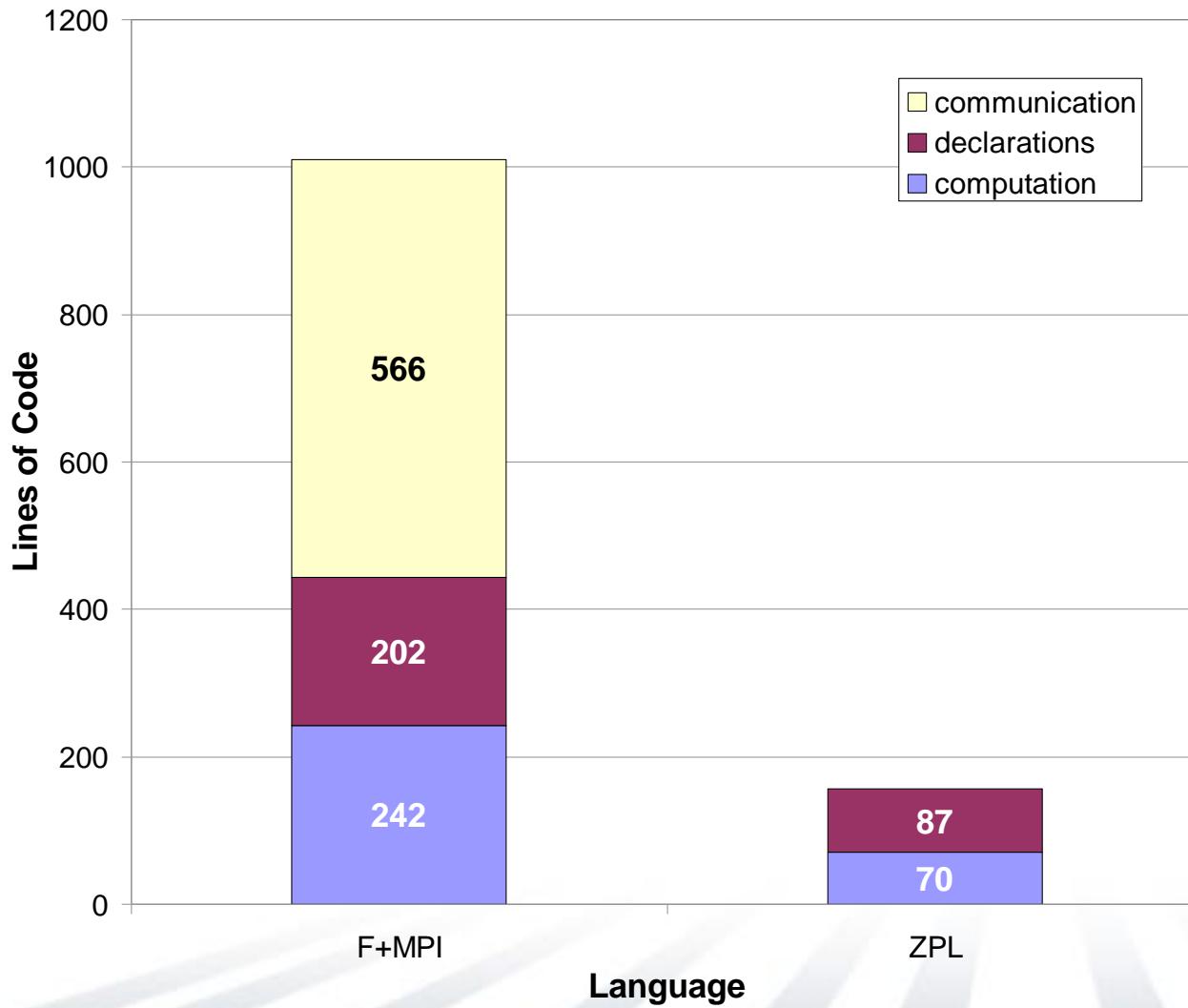
NAS MG Speedup: ZPL vs. Fortran + MPI



Generality Notes

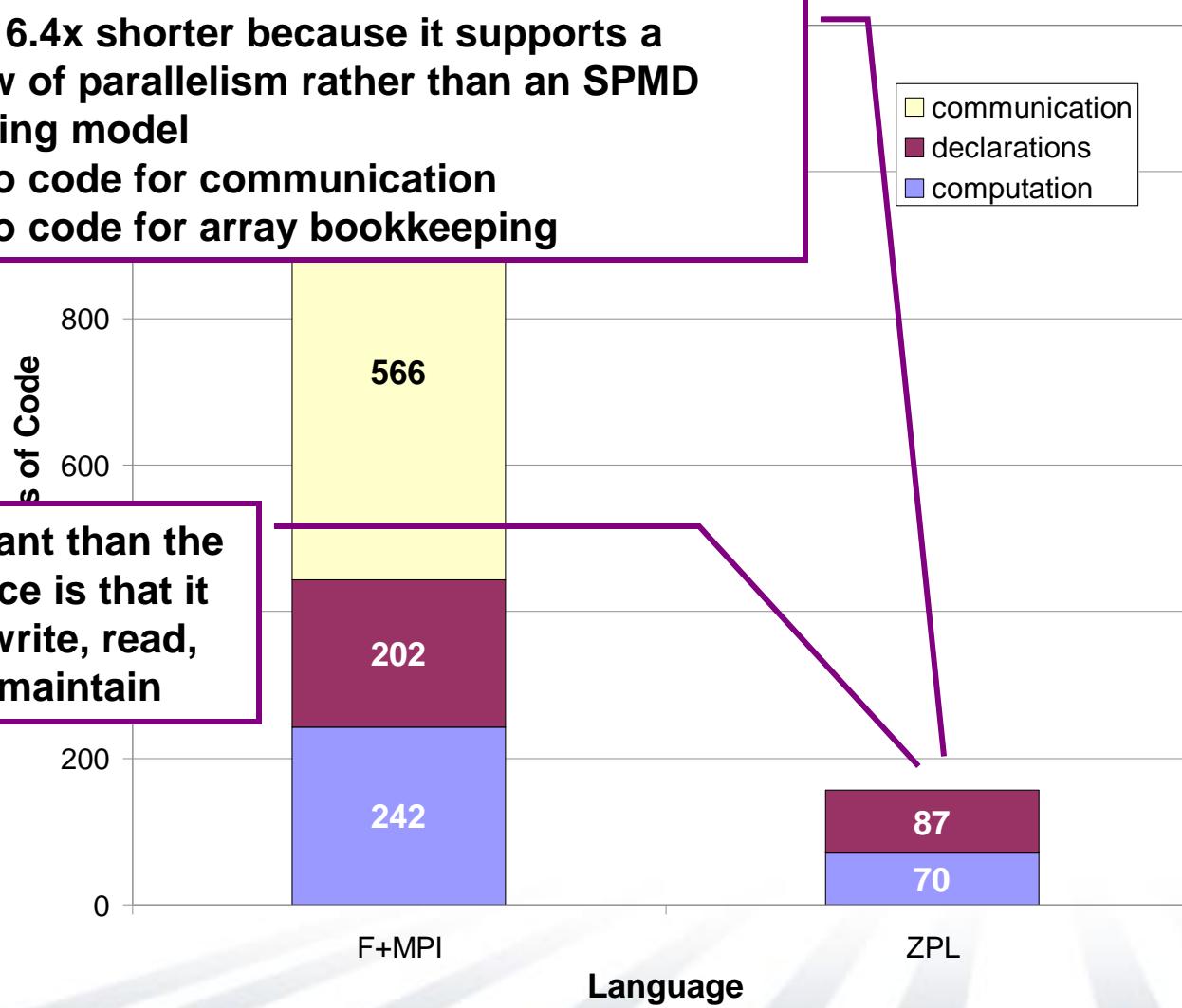


Code Size Notes



Code Size Notes

- the ZPL is 6.4x shorter because it supports a global view of parallelism rather than an SPMD programming model
 - ⇒ little/no code for communication
 - ⇒ little/no code for array bookkeeping



Summarizing Fragmented/SPMD Models

■ Advantages:

- fairly straightforward model of execution
- relatively easy to implement
- reasonable performance on commodity architectures
- portable/ubiquitous
- lots of important scientific work has been accomplished with 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”

Current HPC Programming Notations

■ communication libraries:

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

(fragmented, typically SPMD)
(SPMD)

■ shared memory models:

- OpenMP, pthreads

(global-view, trivially)

■ PGAS languages:

- Co-Array Fortran
- UPC
- Titanium

(SPMD)
(SPMD)
(SPMD)

Scalability Limiters in HPC Programming Models

- Restricted programming and execution models
 - limits applicability to multiple levels of parallelism in HW and SW
- Exposure of low-level implementation mechanisms
 - exposes too much about implementation semantics
 - e.g., saying “how” data should be communicated rather than simply “what” and possibly “when”
 - binds language too tightly to a particular implementation in hardware
 - e.g., MPI for inter-node parallelism +
 - OpenMP/pthreads for inter-core parallelism +
 - directives for intra-core parallelism
- Lack of programmability
 - to a large degree, resulting from the previous two bullets
 - limits ability to modify code – a necessity to tune for these scales

Outline

- ✓ Motivation for Chapel
- ✓ Global-view Programming Models and Scalability
- Language Overview
 - Base Language
 - Parallel Features
 - Task Parallel
 - Data Parallel
 - Locality Features
- Wrap-up

Chapel 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

Base Language: Overview

■ Syntax

- adopt C family of syntax whenever possible/useful
- main departures: declarations/casts, generics, for loops

■ 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

- iterators (in the CLU/Ruby sense, not C++/Java)
- tuples
- latent types / simple static type inference
- rich compile-time language
- configuration variables

Task Parallelism: Task Creation

begin: creates a task for future evaluation

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

sync: waits on all begins created within a dynamic scope

```
sync {  
    begin recursiveTreeSearch(root);  
}
```

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$.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 f() has completed
```

atomic sections: support transactions against memory

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

Task Parallelism: Structured Task Creation

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
```

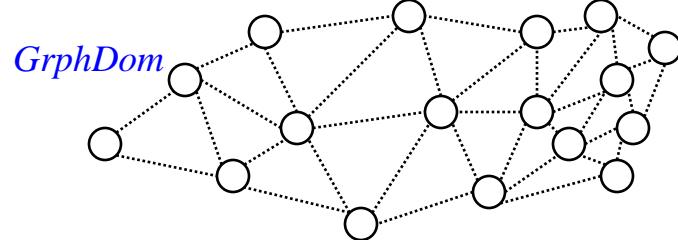
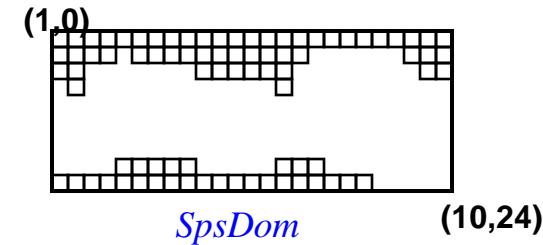
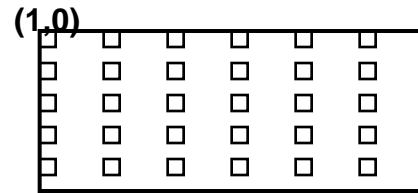
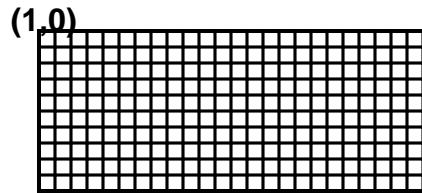
coforall: creates a task per loop iteration

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

Data Parallelism: Domains

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

...integer tuples...



NameDom

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

...anonymous...

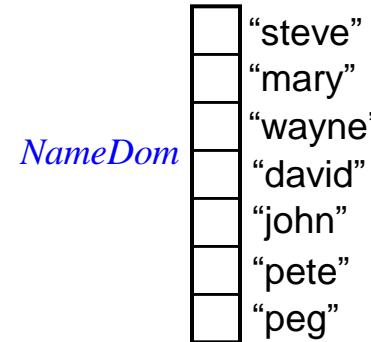
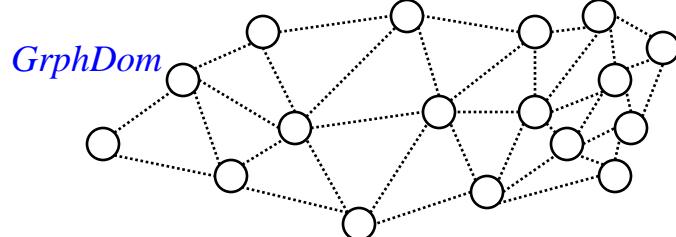
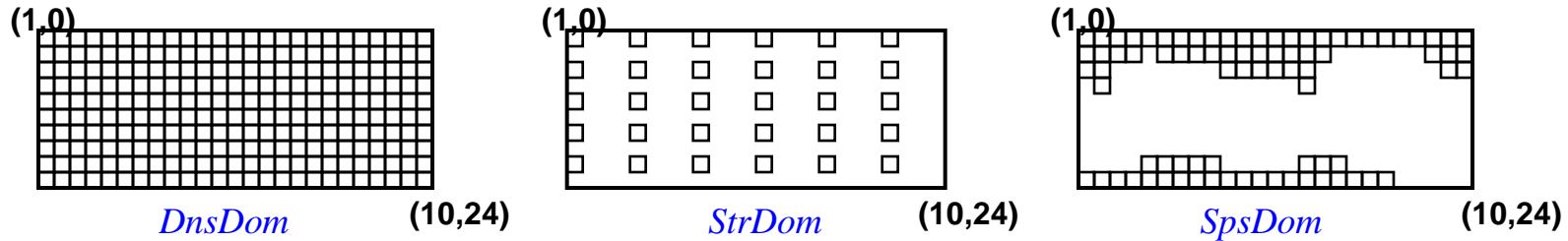
...or arbitrary values.

Data Parallelism: Domain Declarations

```
var DnsDom: domain(2) = [1..10, 0..24],  

    StrDom: subdomain(DnsDom) = DnsDom by (2,4),  

    SpsDom: subdomain(DnsDom) = genIndices();
```



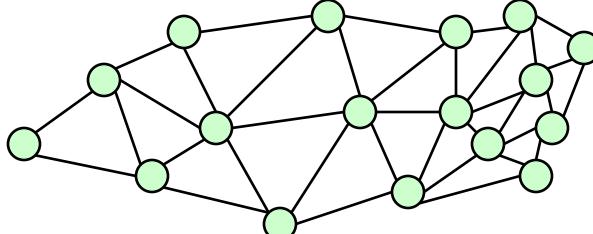
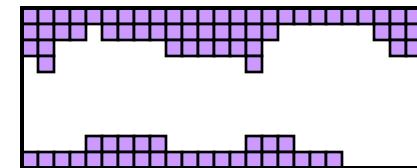
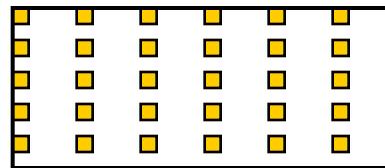
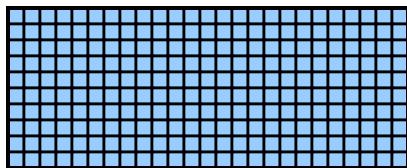
```
var GrphDom: domain(opaque),  

    NameDom: domain(string) = readNames();
```

Data Parallelism: Domains and Arrays

Domains are used to declare arrays...

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

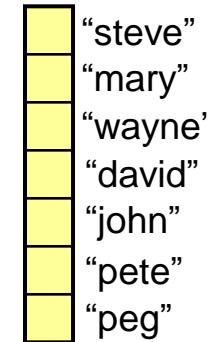
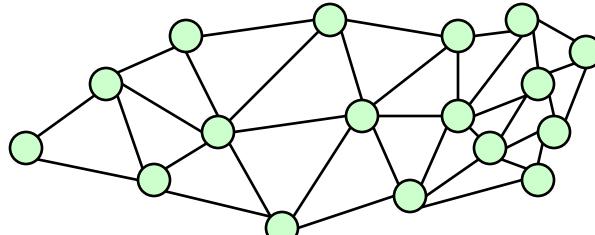
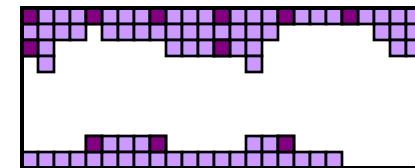
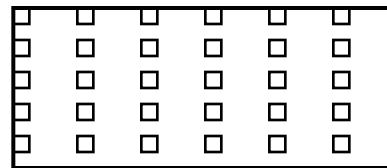
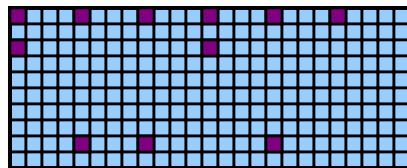


"steve"
"mary"
"wayne"
"david"
"john"
"pete"
"peg"

Data Parallelism: Domain Iteration

...to iterate over index spaces...

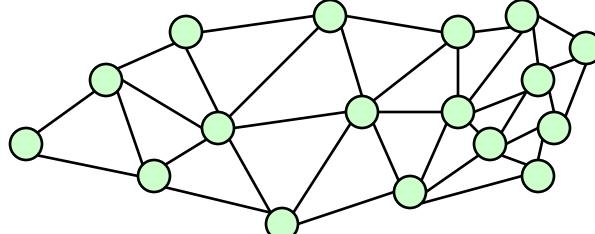
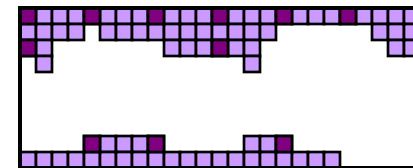
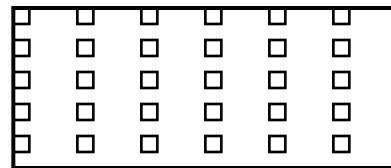
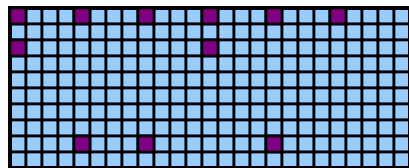
```
forall ij in StrDom {  
    DnsArr(ij) += SpsArr(ij);  
}
```



Data Parallelism: Array Slicing

...to slice arrays...

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

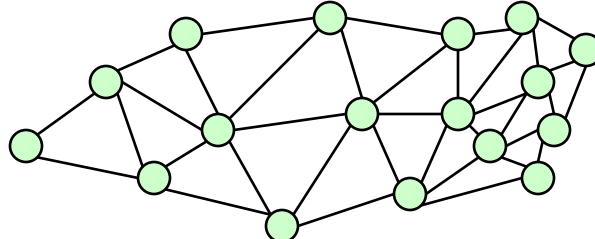
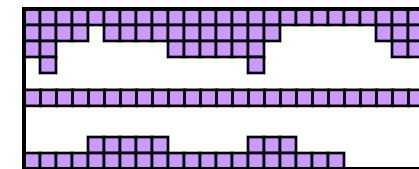
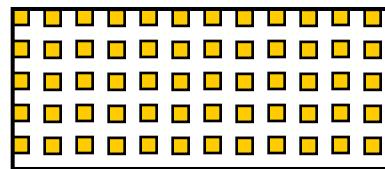
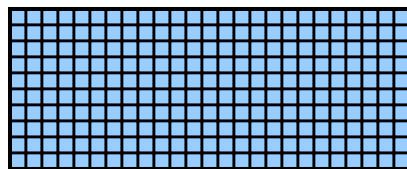


- “steve”
- “mary”
- “wayne”
- “david”
- “john”
- “pete”
- “peg”

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 locality
 - has capacity for storage and processing
 - threads within a locale have ~uniform access to local memory
 - memory within other locales is accessible, but at a price
 - e.g., a multicore processor or SMP node

Locality: Locales

- user specifies # locales on executable command-line

```
prompt> myChapelProg -n1=8
```

- Chapel programs have built-in locale variables:

```
config const numLocales: int;  
const LocaleSpace = [0..numLocales-1],  
    Locales: [LocaleSpace] locale;
```

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

- Programmers can create their own locale views:

```
var CompGrid = Locales.reshape([1..GridRows,  
                                1..GridCols]);
```

0	1	2	3
4	5	6	7

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

0	1
---	---

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

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

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

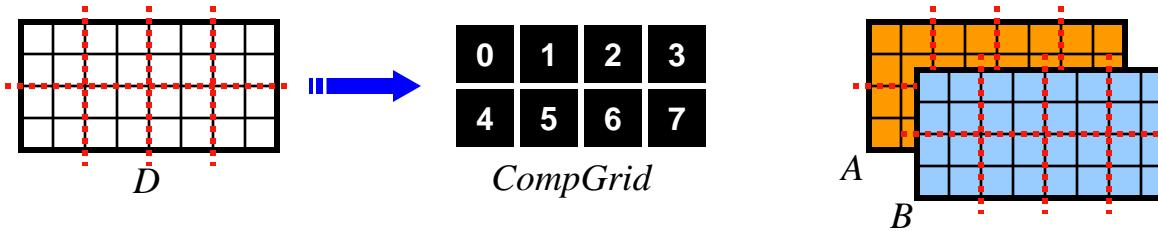
```
cobegin {  
    on TaskALocs do computeTaskA(...);  
    on TaskBLocs do computeTaskB(...);  
    on Locales(0) do computeTaskC(...);  
}
```



Locality: Domain Distribution

Domains may be distributed across locales

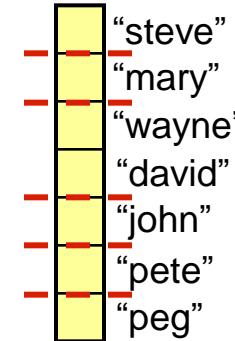
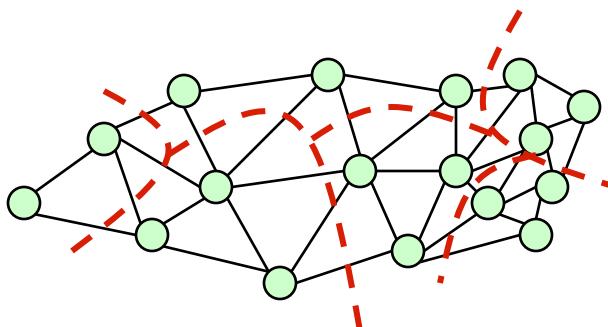
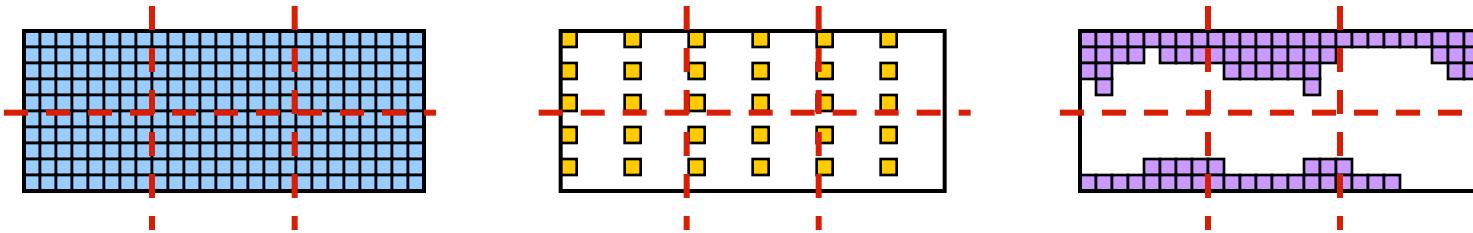
```
var D: domain(2) distributed Block on CompGrid = ...;
```



Locality: Domain Distributions

Distributions specify...

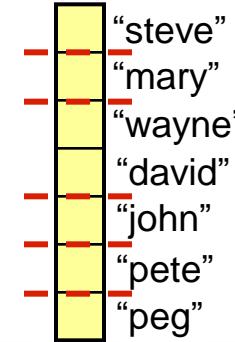
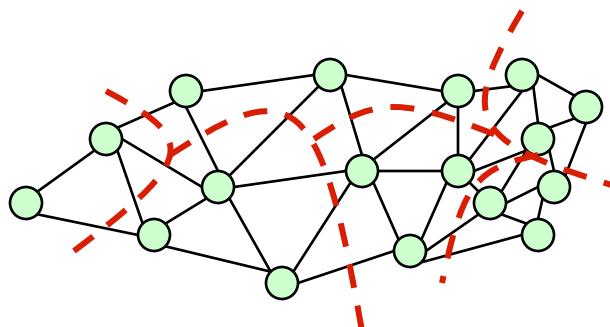
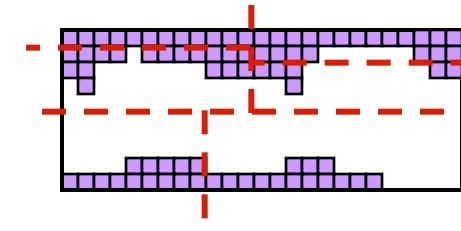
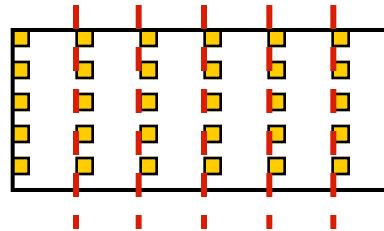
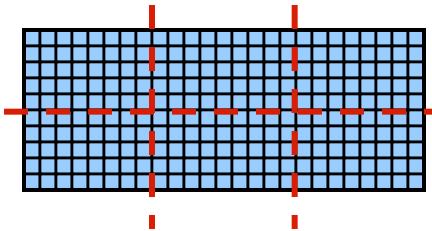
- ...a mapping of indices to locales
- ...per-locale storage for domain indices and array elements
- ...the implementation of parallel operations on domains/arrays



Locality: Domain Distributions

Distributions specify...

- ...a mapping of indices to locales
- ...per-locale storage for domain indices and array elements
- ...the implementation of parallel operations on domains/arrays



Locality: Distributions Overview

Distributions: “recipes for distributed arrays”

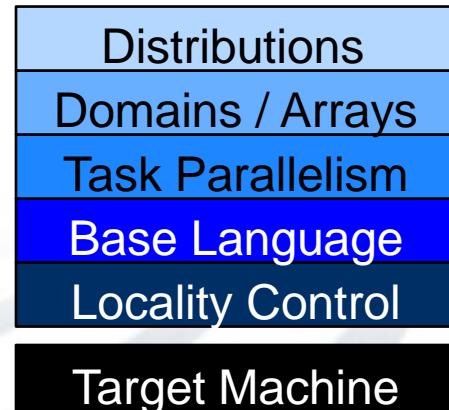
- Intuitively, distributions support the lowering...
 - ...from: the user’s global view of a distributed array
 - ...to: the fragmented implementation on a distributed memory machine
- Users can implement custom distributions:
 - written using task parallel features, on clauses, domains/arrays
 - must implement standard interface:
 - **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

Multiresolution Language Design

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

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



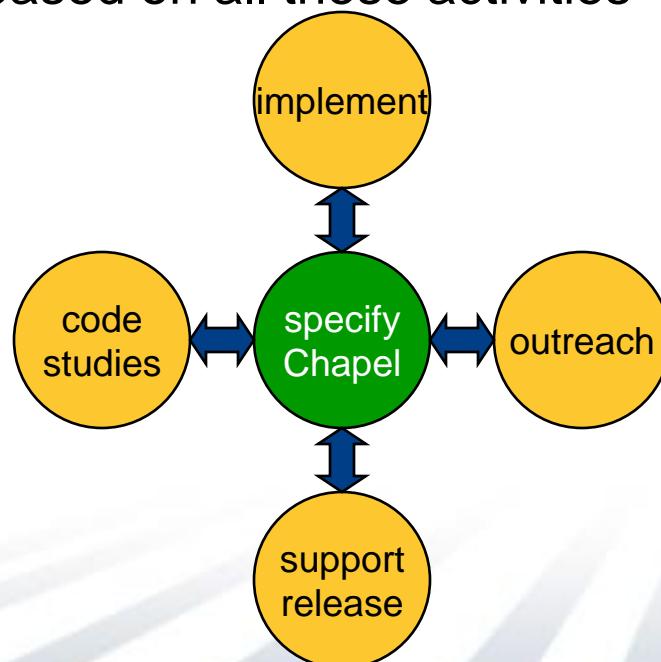
Outline

- ✓ Motivation for Chapel
- ✓ Global-view Programming Models and Scalability
- ✓ Language Overview
- Wrap-up

Chapel Work

- Chapel Team's Focus:

- specify Chapel syntax and semantics
- implement prototype compiler for Chapel
- support users of preliminary releases
- code studies of benchmarks, applications, and libraries in Chapel
- community outreach to inform and learn from users/researchers
- 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
- Early releases to ~40 users at ~20 sites (academic, gov't, industry)

HPC vs. Datacenter concerns

Concern	HPC	Datacenter*
Scalability	Crucial	Crucial
Locality/affinity	Crucial for performance	Crucial for performance
Portability to future technologies	Important	Important
Power/cooling	Increasingly an issue	Increasingly an issue
Flavor of Parallelism	Lots of data dependence, communication	Pleasingly parallel + reductions
Reliability/robustness	Checkpoint/restart	Redundancy + dynamic monitoring
Data types	Floating point (historically)	Strings, integers
Data structures	Multidimensional arrays / unstructured graphs	???
Memory Use	Lots (but generally in-core)	Lots (& generally out-of-core)

* Based on my rather limited understanding...

Chapel for Datacenter Computations?

Some appropriate features:

- large, distributed data structures (“arrays”)
- application of scalar functions to arrays
- reductions: standard and user-defined
- ability to reason about locality, machine resources
- abstraction away from implementing mechanisms
- designed for generality

Yet also some areas requiring innovation/research:

- language-level support for redundancy/reliability?
- extend domains and distributions to out-of-core computations?
- interpreted Chapel for interactive data exploration?
- your ideas here...

A potentially interesting collaboration?

(We’re open to others as well...)

Summary

*Programming languages can help with scalability,
given appropriate design and abstractions*

- Abstractions must map well to hardware capabilities
 - capable of resulting in good performance
 - avoid encoding more about hardware than necessary
- Should support ability to drop to lower levels when required
- Should support ability to control and reason about locality
- Must be as general-purpose as target community requires

Chapel Team



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Questions?

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