Chapel: Global HPCC Benchmarks and Status Update

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Chapel

Chapel: a new parallel language being developed by Cray

- Themes:
 - general parallelism
 - data-, task-, nested parallelism using global-view abstractions
 - general parallel architectures
 - locality control
 - data distribution
 - task placement (typically data-driven)
 - narrow gap between mainstream and parallel languages
 - object-oriented programming (OOP)
 - type inference and generic programming



Chapel's Setting: HPCS

- HPCS: High Productivity Computing Systems
 - Goal: Raise productivity by 10× for the year 2010
 - Productivity = Performance
 - + Programmability
 - + Portability
 - + Robustness
- Phase II: Cray, IBM, Sun (July 2003 June 2006)
 - Evaluation of the entire system architecture's impact on productivity...
 - processors, memory, network, I/O, OS, runtime, compilers, tools, ...
 - ...and new languages:
 - IBM: X10

Sun: Fortress

Cray: Chapel

- Phase III: Cray, IBM (July 2006 2010)
 - Implement the systems and technologies resulting from phase II



Chapel and Productivity

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

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Outline

- √ Chapel Overview
- > HPC Challenge Benchmarks in Chapel
 - STREAM Triad
 - Random Access
 - 1D FFT
- □ Project Status and User Activities



HPC Challenge Overview

Motivation: Growing realization that top-500 often fails to reflect practical/sustained performance

- measured using HPL, which essentially measures peak FLOP rate
- user applications often constrained by memory, network, ...

HPC Challenge (HPCC):

- suite of 7 benchmarks to measure various system characteristics
- annual competition based on 4 of the HPCC benchmarks
 - class 1: best performance (award per benchmark)
 - class 2: most productive
 - 50% performance
 - 50% code elegance, size, clarity

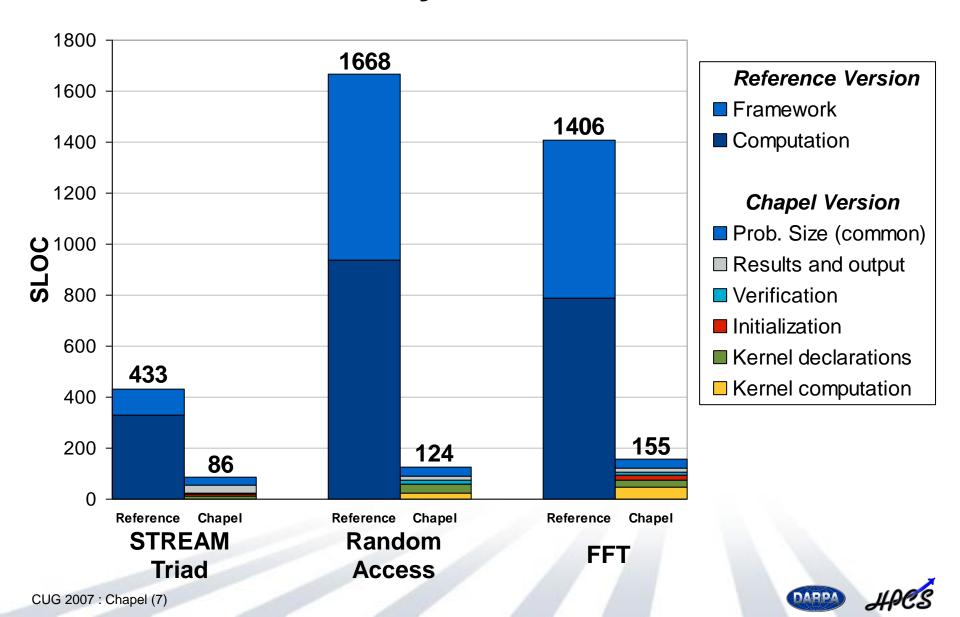
For more information:

- HPCC Benchmarks: http://icl.cs.utk.edu/hpcc/
- HPCC Competition: http://www.hpcchallenge.org

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Code Size Summary



STREAM Triad







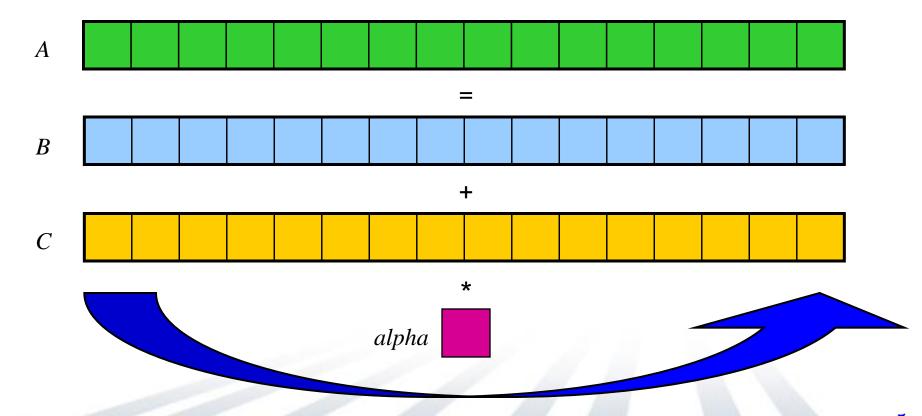


Introduction to STREAM Triad

Given: m-element vectors A, B, C

Compute: $\forall i \in 1..m, A_i = B_i + \alpha \cdot C_i$

Pictorially:



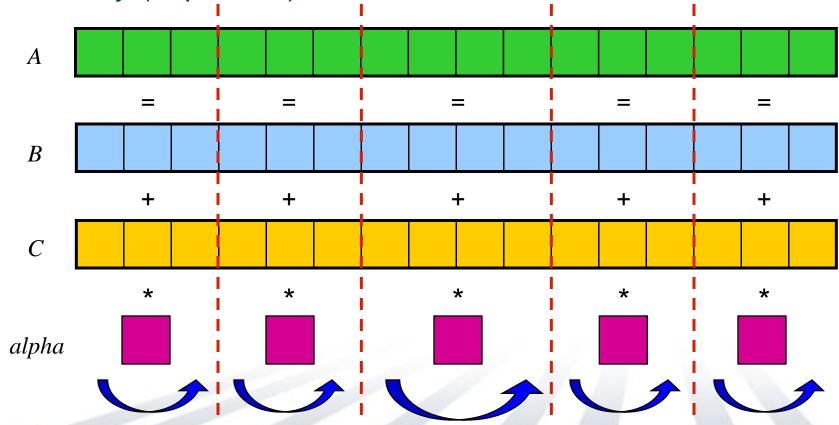


Introduction to STREAM Triad

Given: m-element vectors A, B, C

Compute: $\forall i \in 1..m, A_i = B_i + \alpha \cdot C_i$

Pictorially (in parallel):



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STREAM Triad: Some Declarations



STREAM Triad: Some Declarations

```
const m = computeProblemSize(elemType, numVectors),
      alpha = 3.0;
```

Chapel Variable Declarations

```
{ var | const | param } <name> [: <definition>] [= <initializer>]
    var \Rightarrow can change values
    const ⇒ a run-time constant (can't change values after initialization)
    param ⇒ a compile-time constant
```

May omit definition or initializer, but not both If definition omitted, type inferred from initializer If initializer omitted, variable initialized using type's default value

Here, m has no definition, so its type is inferred using the return type of computeProblemSize() -- an int

Similarly, *alpha* is inferred to be a real floating point value



STREAM Triad: Some Declarations

config const m = computeProblemSize(elemType, numVectors), alpha = 3.0;

Configuration Variables

Preceding a variable declaration with **config** allows it to be initialized on the command-line, overriding its default initializer

config const/var ⇒ can be overridden on executable command-line **config param** \Rightarrow can be overridden on compiler command-line

prompt> stream --m=10000 --alpha=3.14159265





```
const ProblemSpace: domain(1) distributed(Block) = [1..m];
var A, B, C: [ProblemSpace] elemType;
```

```
A = B + alpha * C;
```



const ProblemSpace: domain(1) distributed(Block) = [1..m]; C: [ProblemSpace] elemType: **Declare a domain** domain: a first-class index set, potentially distributed (think of it as the size and shape of an array) $domain(1) \Rightarrow 1D$ arithmetic domain, indices are integers $[1..m] \Rightarrow$ a 1D arithmetic domain literal defining the index set: {1, 2, ..., *m*} **ProblemSpace** m

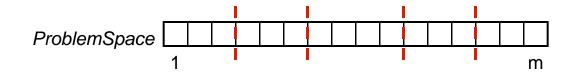


```
const ProblemSpace: domain(1) distributed(Block) = [1..m];
var A, B, C: [ProblemSpace] elemType;
```

Specify the domain's distribution

distribution: describes how to map the domain indices to locales, and how to implement domains (and their arrays)

distributed(Block) ⇒ break the indices into *numLocales* consecutive blocks



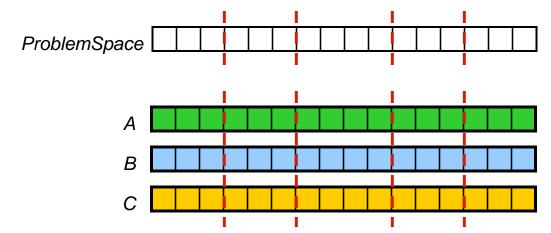


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var A, B, C: [ProblemSpace] elemType;
```

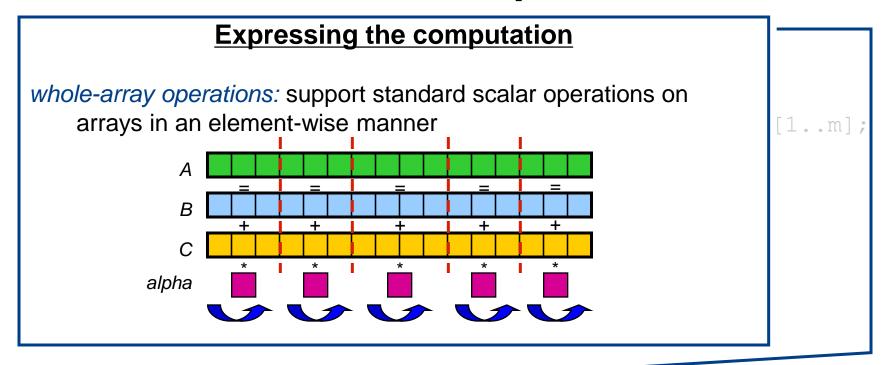
Declare arrays

arrays: mappings from domains (index sets) to variables. Several flavors:

- dense and sparse rectilinear (indexed by integer tuples)
- associative arrays (indexed by value types)
- opaque arrays (indexed anonymously to represent sets & graphs)







$$A = B + alpha * C;$$



```
const ProblemSpace: domain(1) distributed(Block) = [1..m];
var A, B, C: [ProblemSpace] elemType;
```

```
A = B + alpha * C;
```

Random Access





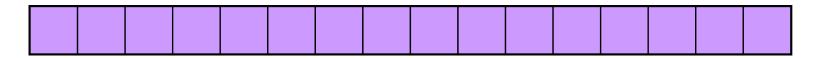




Given: *m*-element table *T* (where $m = 2^n$ and initially $T_i = i$)

Compute: N_U random updates to the table using bitwise-xor

Pictorially:



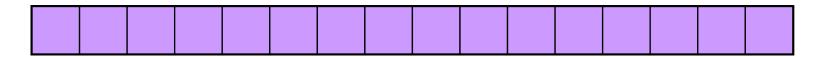


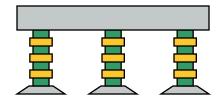


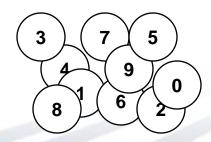
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Pictorially:







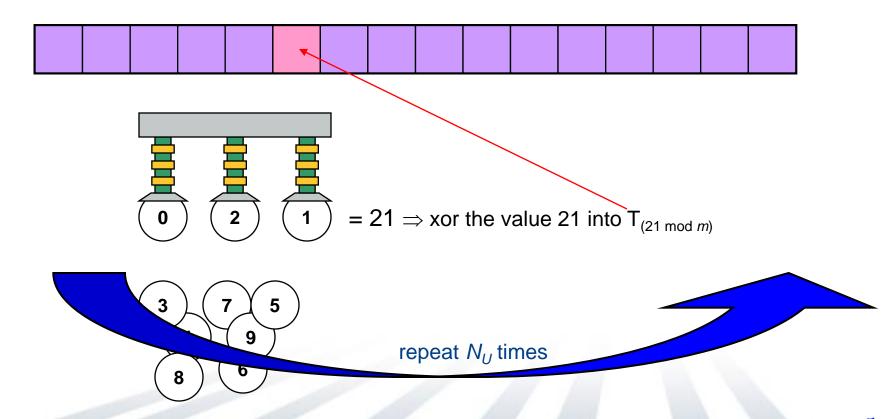


Given: m-element table T (where $m = 2^n$ and initially $T_i = i$)

Compute: N_U random updates to the table using bitwise-xor

Pictorially:

CUG 2007 : Chapel (23)

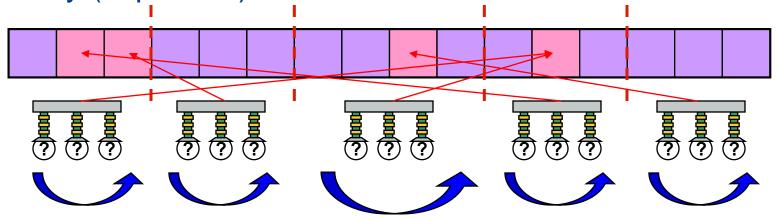




Given: m-element table T (where $m = 2^n$ and initially $T_i = i$)

Compute: N_U random updates to the table using bitwise-xor

Pictorially (in parallel):

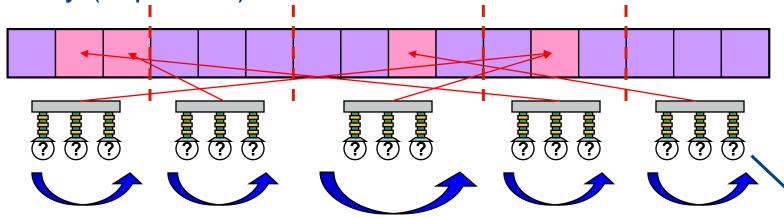




Given: m-element table T (where $m=2^n$ and initially $T_i=i$)

Compute: N_{IJ} random updates to the table using bitwise-xor

Pictorially (in parallel):



Random Numbers

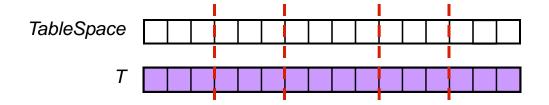
Not actually generated using lotto ping-pong balls! Instead, implement a pseudo-random stream:

- kth random value can be generated at some cost
- given the kth random value, can generate the (k+1)-st much more cheaply



Random Access: Domains and Arrays

```
const TableSpace: domain(1) distributed(Block) = [0..m);
var T: [TableSpace] elemType;
const UpdateSpace: domain(1) distributed(Block) = [0..N U);
```







Random Access: Random Value Iterator

```
iterator RAStream(block) {
  var val = getNthRandom(block.low);
  for i in block {
    getNextRandom(val);
   yield val;
def getNthRandom(in n) { ... }
def getNextRandom(inout x) { ... }
```



Random Access: Random Value Iterator

```
iterator RAStream(block) {
  var val = getNthRandom(block.low);
  for i in block {
     getNextRandom(val);
     yield val;
                             Defining an iterator
     iterator: similar to a function but generates a stream of return values;
       invoked using for and forall loops
def
     yield: like a return statement but the iterator's execution continues
       logically after returning the value
def
     RAStream(): an iterator that generates a random value for each index in
       block
     e.g., to iterate over the entire stream sequentially, one could use:
         for r in RAStream([0..N U)) { ... }
```



Random Access: Random Value Iterator

```
iterator RAStream(block) {
  var val = getNthRandom(block.low);
  for i in block {
    getNextRandom(val);
   yield val;
def getNthRandom(in n) { ... }
def getNextRandom(inout x) { ... }
```



Random Access: Computation

```
[i in TableSpace] T(i) = i;

forall block in UpdateSpace.subBlocks do
  for r in RAStream(block) do
    T(r & indexMask) ^= r;
```



Random Access: Computation

```
[i in TableSpace] T(i) = i;
```

Initialization

Uses *forall expression* to initialize table

```
forall block in UpdateSpace.subBlocks do
for r in RAStream(block) do
   T(r & indexMask) ^= r;
```

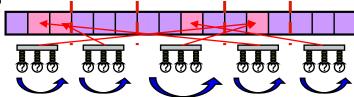
Computing the Updates

Express table updates by invoking iterators:

subBlocks: a standard iterator that generates blocks of indices appropriate for the target machine's parallelism

RAStream(): our iterator for generating random values

Effectively: generate parallel chunks of work; iterate over chunks serially performing updates





Random Access: Computation

```
[i in TableSpace] T(i) = i;

forall block in UpdateSpace.subBlocks do
   for r in RAStream(block) do
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```

FFT







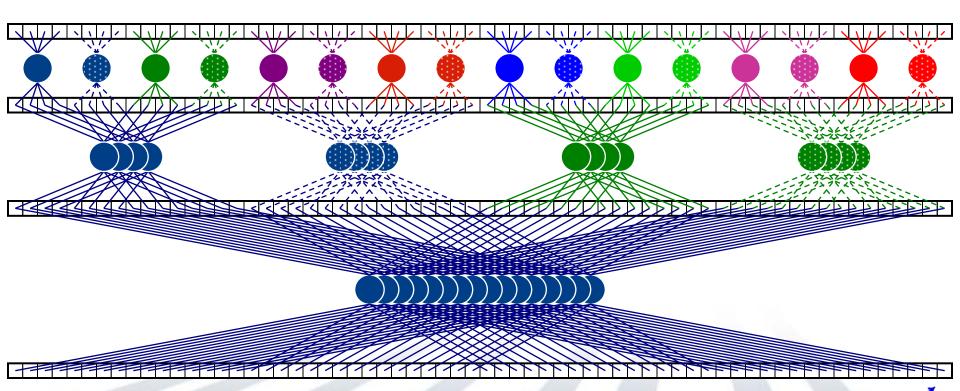


Introduction to FFT

Given: m-element vector z of complex numbers (where $m = 2^n$)

Compute: 1D Discrete Fourier Transform of z

Pictorially (using a radix-4 algorithm):







FFT: Computation

```
for i in [2..log2(numElements)) by 2 {
  const m = radix*span, m2 = 2*m;
  forall (k, k1) in (Adom by m2, 0...) {
    var wk2 = ..., wk1 = ..., wk3 = ...;
    forall j in [k..k+span] do
      butterfly(wk1, wk2, wk3, A[j...j+3*span by span]);
    wk1 = ...; wk3 = ...; wk2 *= 1.0i;
    forall j in [k+m..k+m+span) do
      butterfly(wk1, wk2, wk3, A[j..j+3*span by span]);
  span *= radix;
def butterfly(wk1, wk2, wk3, inout A:[1..radix]) { ... }
```



FFT: Computation

Sequential loop to express phases of computation

```
for i in [2..log2(numElements)) by 2 {
  const m = radix*span, m2 = 2*m;
  forall (k, k1) in (Adom by m2, 0...) {
    var wk2 = ..., wk1 = ..., wk3 = ...;
                                       Nested forall loops to express
                                        a phase's parallel butterflies
    forall j in [k..k+span) do
      butterfly(wk1, wk2, wk3, A[j..j+3*span by span]);
                                        Support for complex and
    wk1 = ...; wk3 = ...; wk2 *= 1.0i;
                                         imaginary types simplifies math
    forall j in [k+m..k+m+span) do
      butterfly(wk1, wk2, wk3, A[j..j+3*span by span]);
                                 Generic arguments allow butterfly() to be
  span *= radix;
                                 called with complex, real, or imaginary
                                 twiddle factors
def butterfly(wk1, wk2, wk3, inout A:[1..radix]) { ... }
```



FFT: Computation

```
for i in [2..log2(numElements)) by 2 {
  const m = radix*span, m2 = 2*m;
  forall (k, k1) in (Adom by m2, 0...) {
    var wk2 = ..., wk1 = ..., wk3 = ...;
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      butterfly(wk1, wk2, wk3, A[j...j+3*span by span]);
    wk1 = ...; wk3 = ...; wk2 *= 1.0i;
    forall j in [k+m..k+m+span) do
      butterfly(wk1, wk2, wk3, A[j..j+3*span by span]);
  span *= radix;
def butterfly(wk1, wk2, wk3, inout A:[1..radix]) { ... }
```



HPCC Status, Next Steps

HPCC Status:

- all codes compile and run today
- current compiler only targets a single node
- serial performance approaching hand-coded C on a daily basis
- CUG paper...
 - ...contains full source listings
 - ...covers codes in more detail
 - ...describes performance advantages and challenges in Chapel

What's Next?

- demonstrate performance for these codes
 - continue optimizing serial performance
 - add compiler support for targeting multiple nodes
- finish implementing HPL



HPCC Summary

- Chapel supports HPCC codes attractively
 - clear, concise, general
 - parallelism expressed in architecturally-neutral way
 - benefit from Chapel's global-view parallelism
 - utilizes generic programming and modern SW Engineering principles
 - should serve as an excellent reference for future HPCC competitors
- Note that HPCC benchmarks are relatively simple
 - all data structures are 1D vectors
 - locality very data driven
 - minimal task- & nested parallelism
 - little need for OOP, abstraction

...HPCC designed to stress systems, not languages

• would like to see similar competitions emerge for richer computations



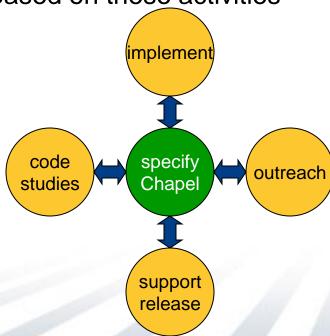
Outline

- √ Chapel Overview
- ✓ HPC Challenge Benchmarks in Chapel
 - **✓** STREAM Triad
 - ✓ Random Access
 - ✓1D FFT
- Project Status and User Activities



Chapel Work

- Chapel Team's Focus:
 - specify Chapel syntax and semantics
 - implement prototype Chapel compiler
 - code studies of benchmarks, applications, and libraries in Chapel
 - community outreach to inform and learn from users
 - support users evaluating the language
 - refine language based on these activities



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Project Status, Next Steps

Chapel specification:

- revised draft language specification available on Chapel website
- editing to add additional examples & rationale; improve clarity

Compiler implementation:

- improving serial performance
- starting on distributed memory implementation
- adding missing serial features

Code studies:

- NAS Parallel Benchmarks: CG (well underway), IS, FT, MG
- Linear Algebra routines: block LU, block Cholesky, matrix mult.
- Other applications of interest: Fast Multipole Method, SSCA2, ...

Release:

- made a preliminary release to government team December 2006
- initial response from those users has been positive, encouraging
- next release due Summer 2007

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Notable User Studies

- Two main efforts to date, both at ORNL:
 - Robert Harrison, Wael Elwasif, David Bernholdt, Aniruddha Shet
 - Fock matrix computations using producer-consumer parallelism
 - coupled model idioms (e.g., for use in CCSM)
 - Richard Barrett, Stephen Poole, Philip Roth
 - stencil idioms: 2D, 3D, sparse
 - sweep3D & wavefront-style computations
- In both cases...
 - ...great technical discussions and feedback
 - ...valuable sanity-check for language and implementation
 - ...studies comparing with Fortress, X10 forthcoming



Chapel Contributors

Current:

- Brad Chamberlain
- Steven Deitz
- Mary Beth Hribar
- David Iten
- (Your name here? We're hiring...)

Alumni:

- David Callahan
- Hans Zima (CalTech/JPL)
- John Plevyak
- Wayne Wong
- Shannon Hoffswell
- Roxana Diaconescu (CalTech)
- Mark James (JPL)
- Mackale Joyner (2005 intern, Rice University)
- Robert Bocchino (2006 intern, UIUC)

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For More Information...

BOF today at 4pm

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http://chapel.cs.washington.edu

Your feedback desired!