Targeting GPUs and Other Hierarchical Architectures in Chapel

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Goals of this work

- Leverage Chapel to generate efficient data parallel code to execute on GPUs
- Provide the ability to support execution on any hierarchical parallel architecture
- Generalize parallel loops for efficiency and portability across many parallel architectures and runtimes



A = alpha * B + C

MOTIVATING EXAMPLE: HPC CHALLENGE (HPCC) STREAM TRIAD



```
config const m = 1000;

const alpha = 3.0;

const ProbSpace = [1..m];

var A, B, C: [ProbSpace] real;

forall (a,b,c) in (A,B,C) do
  a = b + alpha * c;

By default, executes on a multicore
```



```
config const m = 1000;
const alpha = 3.0;

const ProbSpace = [1..m] dmapped GPUDist(rank=1);

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No changes required to the computation for other architectures
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No need for explicit transfers of data between host and device (e.g. cudaMemcpy(...))

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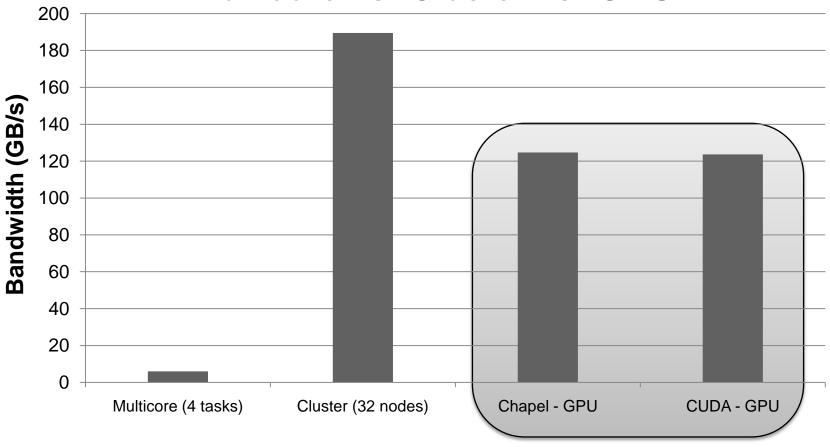
STREAM Triad (current practice)

```
#define N
                2000000
int main() {
                                   CUDA
  float *d a, *d b, *d c;
  float scalar:
  cudaMalloc((void**)&d a, sizeof(float)*N);
  cudaMalloc((void**)&d b, sizeof(float)*N);
  cudaMalloc((void**)&d c, sizeof(float)*N);
  dim3 dimBlock(128);
  dim3 dimGrid(N/dimBlock.x);
  if ( N % dimBlock.x != 0 ) dimGrid.x+=1;
  set array<<<dimGrid,dimBlock>>>(d b, .5f, N);
  set array<<<dimGrid,dimBlock>>>(d c, .5f, N);
  scalar=3.0f;
  STREAM Triad<<dimGrid,dimBlock>>>(d b, d c, d a, scalar, N);
  cudaThreadSynchronize();
  cudaFree(d a);
  cudaFree(d b);
  cudaFree(d c);
global void set array(float *a, float value, int len) {
  int idx = threadIdx.x + blockIdx.x * blockDim.x;
  if (idx < len) a[idx] = value;
global void STREAM Triad( float *a, float *b, float *c,
                              float scalar, int len) {
  int idx = threadIdx.x + blockIdx.x * blockDim.x;
  if (idx < len) c[idx] = a[idx]+scalar*b[idx];</pre>
```

```
#include <hpcc.h>
#ifdef OPENMP
#include <omp.h>
                                    MPI + OpenMP
#endif
static int VectorSize;
static double *a, *b, *c;
int HPCC StarStream(HPCC Params *params)
 int myRank, commSize;
 int rv, errCount;
 MPI Comm comm = MPI COMM WORLD;
 MPI Comm size ( comm, &commSize );
 MPI Comm rank ( comm, &myRank );
 rv = HPCC Stream( params, 0 == myRank);
 MPI Reduce( &rv, &errCount, 1, MPI INT, MPI SUM, 0, comm );
 return errCount;
int HPCC Stream (HPCC Params *params, int doIO) {
 register int j;
  double scalar;
  VectorSize = HPCC LocalVectorSize( params, 3, sizeof(double), 0 );
  a = HPCC XMALLOC( double, VectorSize );
  b = HPCC XMALLOC( double, VectorSize );
  c = HPCC XMALLOC( double, VectorSize );
  if (!a || !b || !c) {
   if (c) HPCC free(c);
   if (b) HPCC free(b);
   if (a) HPCC free(a);
   if (doIO) {
     fprintf( outFile, "Failed to allocate memory (%d).\n", VectorSize );
     fclose( outFile );
   return 1;
#ifdef OPENMP
#pragma omp parallel for
 for (j=0; j<VectorSize; j++) {
   b[j] = 2.0;
   c[j] = 0.0;
  scalar = 3.0;
#ifdef OPENMP
#pragma omp parallel for
 for (j=0; j<VectorSize; j++)
  a[j] = b[j]+scalar*c[j];
 HPCC free(c);
 HPCC free (b);
 HPCC free(a);
  return 0;
```



Performance of STREAM Triad Multicore vs. Cluster vs. GPU

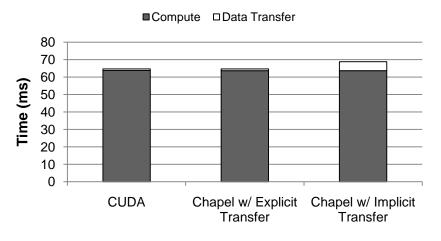




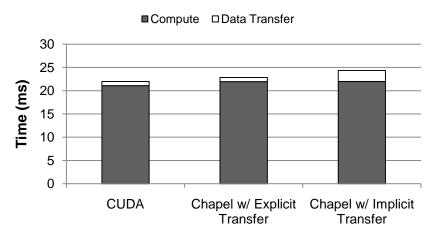


Parboil Benchmarks

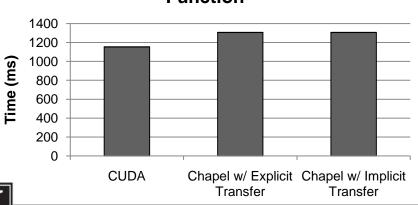




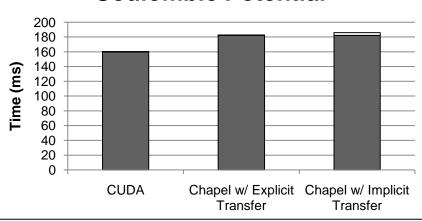
MRI-Q



Two Point Angular Correlation Function



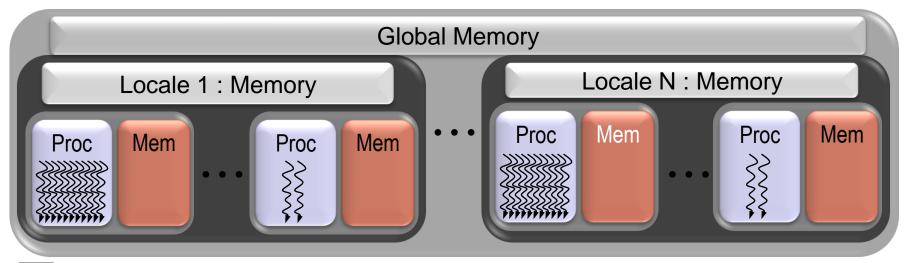
Coulombic Potential





Hierarchical Locales

- Today: Support for locality is horizontal across nodes
 - Lacking support for vertical locality (within a node)
- Challenge: How to map a computation onto a deeply hierarchical parallel architecture



Programming Hierarchical Locales

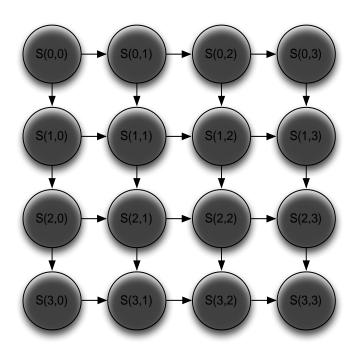
- Three main components:
 - 1. Hierarchical Machine Model
 - Method to define the machine (locale hierarchy)
 - 2. Task Execution Model onto locale (or sublocale)
 - Synchronous (forall loops, whole-array operations)
 - Asynchronous (on, begin, cobegin, coforall, etc.
 - Data Model
 - Distribution of data across (and within) locales



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In progress: Generalizing Parallel Loops

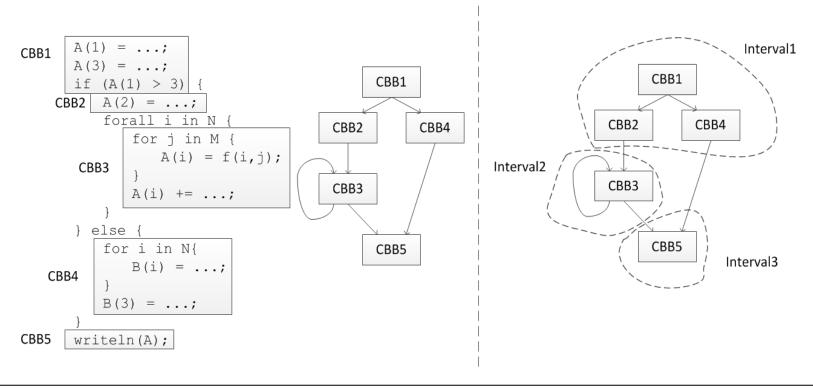
- Method to perform wavefront & pipeline computations as simple loops
- Programmer provides annotated dependence information
- Compiler performs necessary transformations to parallelize the loop





In progress: Compiling for Data Flow

- Transformation to convert original Chapel program into a task graph
 - Nodes represent an atomic unit of computation
 - Edges express dependences between tasks
- Compiler maps the graph onto a data flow runtime and architecture





Any questions : **asideIn2 at illinois.edu**Thanks!

