The resurgence of parallel programming languages

Jamie Hanlon &
Simon McIntosh-Smith
University of Bristol
Microelectronics Research Group
hanlon@cs.bris.ac.uk







The Microelectronics Research Group at the University of Bristol

www.cs.bris.ac.uk/Research/Micro





The team



Simon McIntosh-Smith Head of Group



Prof David May



Prof Dhiraj Pradhan



Dr Jose Nunez-Yanez



Dr Kerstin Eder



Dr Simon Hollis



Dr Dinesh Pamunuwa

7 tenured staff, 6 research assistants, 16 PhD students



Group expertise

Energy Aware COmputing (EACO):

- Multi-core and many-core computer architectures
 - Inmos, XMOS, ClearSpeed, Pixelfusion, ...
- Algorithms for heterogeneous architectures (GPUs, OpenCL)
- Electronic and Optical Network on Chip (NoC)
- Reconfigurable architectures (FPGA)
- Design verification (formal and simulation-based), formal specification and analysis
- Silicon process variation
- Fault tolerant design (hardware and software)
- Design methodologies, modelling & simulation of MNT based structures and systems



Coverview

- Parallelism in computing
- Overview and discussion of two current parallel languages
 - Chapel (HPC)
 - OpenCL (desktop/embedded)
- Moving forward
 - Heterogeneous System Architecture
 - Research into scalable general purpose parallel architectures





Didn't parallel computing use to be a niche?





& A long history in HPC...





We But now parallelism is mainstream

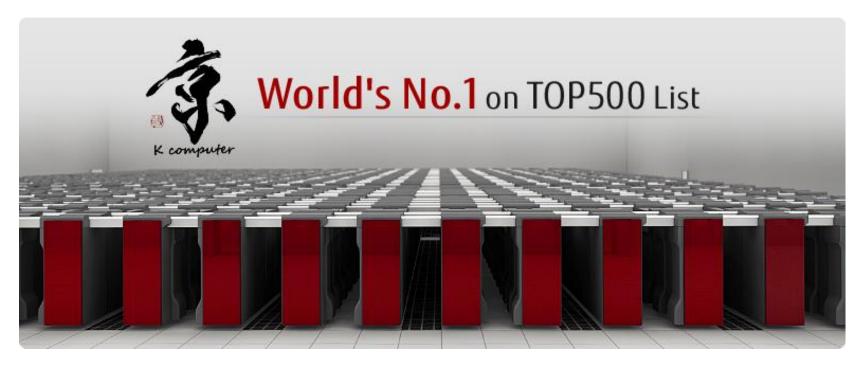


Quad-core ARM Cortex A9 CPU

Quad-core SGX543MP4+ Imagination GPU



KHPC stronger than ever



- 705,024 SPARC64 processor cores delivering 10.51 petaflops (10 quadrillion calculations per second)
- No GPUs or accelerators
- 9.9 MW



Increasing use of GPUs at high end



- Tianhe-1A in Tianjin, China (2nd in Top500)
- 2.6 petaflops
- 14,336 Intel 2.93 GHz CPUs (57,334 cores)
- 7,168 NVIDIA Tesla M2050 GPUs (100,000 cores)
- 4 MW power consumption





http://www.nytimes.com/2006/06/14/technology/14search.html

Report: Google Uses About 900,000 Servers (Aug 1st 2011)

http://www.datacenterknowledge.com/archives/2011/08/01/report-google-uses-about-900000-servers/



The Dalles (84)

OREGON

Snake R.

Portland

Salem

A renaissance in parallel programming

CSP

- Erlang
- Occam-pi
- XC

GPGPU

- OpenCL
- CUDA
- HMPP
- OpenACC

Message-passing

MPI



Multi-threaded

- OpenMP
- Cilk
- Go

Object-orientated

- C++ AMP
- CHARM++

PGAS

- Co-array Fortran
- Chapel
- Unified Parallel C
- X10



Chapel





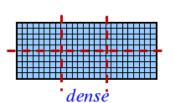
K Chapel

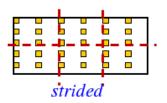
- Cray development funded by DARPA as part of HPCS program
- Partitioned global address space (PGAS) language
 - Central abstraction is a global array partitioned across a system
- Programmer control of locality by allowing explicit affinity of both tasks and data to locales

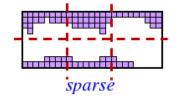


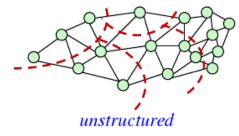
Arrays and distribution

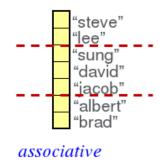
- An array is a general concept and can be declared with different domains
- Can be distributed with a domain map
 - Standard maps and can be user-defined
- Computation can remain the same regardless of a specific distribution













Chapel's data parallelism

Zippered forall:

- loop body sees ith element from each iteration
- Works over:
 - distributed arrays
 - arrays with different distributions
 - user-defined iterators A,B,C could be trees or graphs



MPI+OpenMP

```
#include <hpcc.h>
#ifdef OPENMP
#include <omp.h>
#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
#endif
  for (j=0; j<VectorSize; j++) {</pre>
  b[i] = 2.0;
   c[i] = 0.0;
  scalar = 3.0;
#ifdef OPENMP
#pragma omp parallel for
#endif
  for (j=0; j<VectorSize; j++)</pre>
    a[j] = b[j] + scalar * c[j];
  HPCC free(c);
  HPCC free(b);
  HPCC free(a);
  return 0;
```

Composition in Chapel

Data parallelism

```
cobegin {
  forall (a, b, c) in (A, B, C) do
    a = b + alpha * c;
  forall (d, e, f) in (D, E, F) do
    d = e + beta * f;
}
```

Task parallelism nested in data parallelism

```
forall a in A {
   if a == 0 then
    begin a = f(a)
   else
    a = g(a)
}
```



Issues with Chapel

- HPC-orientated: not suitable for general programming, e.g. embedded platforms
- Locales support only a single level hierarchy
- No load balancing/dynamic resource management
- Too high level? Is it a good abstraction of a parallel machine?





OpenCL(Open Computing Language)



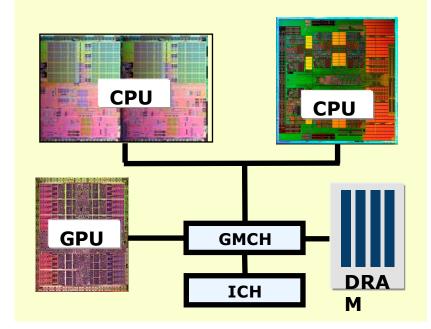


KOpenCL

- Open standard for portable, parallel programming of heterogeneous systems
- Lets programmers write a single portable program that uses all resources in the heterogeneous platform

A modern system includes:

- One or more CPUs
- One or more GPUs
- DSP processors
- ...other devices?

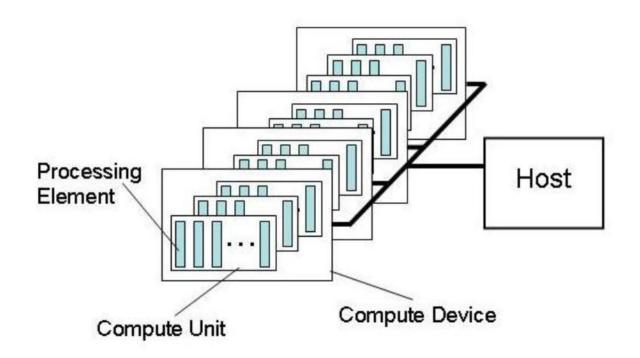


GMCH = graphics memory control hub

ICH = Input/output control hub



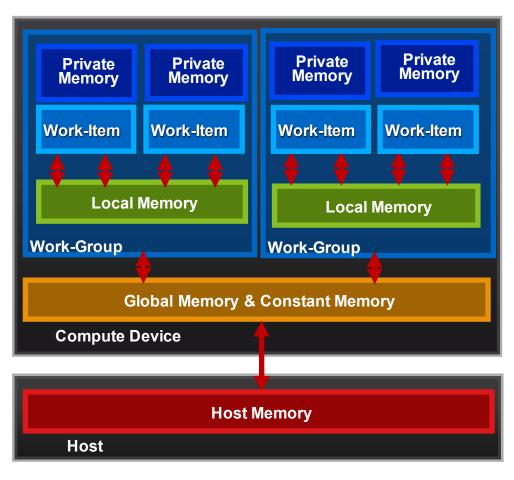
OpenCL platform model



- One <u>Host</u> + one or more <u>Compute Devices</u>
 - Each Compute Device is composed of one or more <u>Compute Units</u>
 - Each Compute Unit is further divided into one or more <u>Processing</u> <u>Elements</u>

OpenCL memory model

- Private Memory
 - Per Work-Item
- Local Memory
 - Shared within a Work-Group
- Global / Constant Memories
 - Visible to all Work-Groups
- Host Memory
 - On the CPU



Memory management is explicit

You must move data from host → global → local *and* back



The BIG idea behind OpenCL

- Replace loops with functions (a <u>kernel</u>) executing at each point in a problem domain (index space).
- E.g., process a 1024 x 1024 image with one kernel invocation per pixel or 1024 x 1024 = 1,048,576 kernel executions

Traditional loops

Data parallel OpenCL



Host program boilerplate

```
// create the OpenCL context on a GPU device
cl context = clCreateContextFromType(0,
   CL DEVICE TYPE GPU, NULL, NULL, NULL);
// get the list of GPU devices associated with context
clGetContextInfo(context, CL CONTEXT DEVICES, 0,
                                          NULL, &cb);
devices = malloc(cb);
clGetContextInfo(context, CL CONTEXT DEVICES, cb,
   devices, NULL);
// create a command-queue
cmd queue = clCreateCommandQueue(context, devices[0],
   \overline{0}, NULL);
// allocate the buffer memory objects
memobjs[0] = clCreateBuffer(context, CL MEM READ ONLY |
   CL MEM COPY HOST PTR, sizeof(cl float)*n, srcA,
   NU\overline{L}L);
memobjs[1] = clCreateBuffer(context,CL MEM READ ONLY |
   CL MEM COPY HOST PTR, sizeof(cl float)*n, srcB,
   NU\overline{L}L);
memobjs[2] = clCreateBuffer(context,CL MEM WRITE ONLY,
                             sizeof(cl float)*n, NULL,
   NULL);
// create the program
program = clCreateProgramWithSource(context, 1,
   &program source, NULL, NULL);
```

```
// build the program
err = clBuildProgram(program, 0, NULL, NULL, NULL,
// create the kernel
kernel = clCreateKernel(program, "vec add", NULL);
// set the args values
err = clSetKernelArg(kernel, 0, (void *) &memobjs[0],
                                 sizeof(cl mem));
err |= clSetKernelArg(kernel, 1, (void *) &memobjs[1],
                                 sizeof(cl mem));
err |= clSetKernelArg(kernel, 2, (void *) &memobjs[2],
                                  sizeof(cl mem));
// set work-item dimensions
global work size[0] = n;
// execute kernel
err = clEnqueueNDRangeKernel(cmd queue, kernel, 1,
   NULL, global work size, NULL, 0, NULL, NULL);
// read output array
err = clEnqueueReadBuffer(context, memobjs[2], CL TRUE,
   0, n*sizeof(cl float), dst, 0, NULL, NULL);
```



Host program boilerplate

```
// create the OpenCL context on a GPU device
cl context = clCreateContextFromType(0,
   CL DEVICE TYPE GPU, NULL, NULL, NULL);
// get the list of GPU devices associated with context
 Define platform and queues
clGetContextInfo(context, CL CONTEXT DEVICES, cb,
   devices, NULL);
// create a command-queue
cmd queue = clCreateCommandQueue(context, devices[0],
  \overline{0}, NULL);
// allocate the buffer memory objects
memobjs[0] = clCreateBuffer(context, CL MEM READ ONLY
   CL MEM COPY HOST PTR, sizeof(cl float)*\overline{n}, src\overline{A},
      Define Memory objects
                                                ONLY |
memo
   NULL);
memobjs[2] = clCreateBuffer(context,CL MEM WRITE ONLY,
                           sizeof(cl float)*n, NULL,
  NULL);
// cre
        Create the program
progra
   &pr
```

```
// build
         Build the program
                                        NULL.
// create the kernel
kernel = clCreateKernel(program, "vec add", NULL);
   Create and setup kernel
err |= clSetKernelArg(kernel, 1, (void *) &memobjs[1],
                             sizeof(cl mem));
err |= clSetKernelArg(kernel, 2, (void *) &memobjs[2],
                              sizeof(cl mem));
// set work-item dimensions
global work size[0] = n;
// execu Execute the kernel
  NULL, global work size, NULL, 0, NULL, NULL);
 Read results back to the host
```



K Issues with OpenCL

- Low level
- It does not compose
 - Disjoint memory address spaces (local/global)
 - Barriers
- It provides no resource management
 - Kernels are a statically allocated resource





Heterogeneous System Architecture (HSA)





K HSA overview

- Announced recently by AMD as new open architecture specification
 - HSAIL virtual ISA
 - HSA memory model
 - HSA dispatch
- Designed to support CPU-GPU integration in APU
- Provides an optimised platform architecture for OpenCL
- Already being adopted by other vendors starting with ARM



KHSA features

- Integration of CPU and GPU in silicon
- Unified address space for CPU and GPU
- Potentially even GPU context switching!
- HSA programming model introduces PGAS-style distributed arrays
 - Memory hierarchy abstraction to address function composition
- First class barrier objects



HSA Intermediate Layer (HSAIL)

- Virtual ISA for parallel programs
- Similar idea to LLVM IR a good target for compilers
- Finalised to specific ISA by a JIT compiler
- Features:
 - Explicitly parallel
 - Support for exceptions, virtual functions and other high-level features
 - Syscall methods (I/O, printf etc.)
 - Debugging support



K HSA memory model

- Compatible with C++11, Java and .NET memory models
- Relaxed consistency



K HSA dispatch

- HSA designed to enable heterogeneous task queuing
 - A work queue per core
 - Distribution of work into queues
 - Load balancing by work stealing



Problems in the long term

- Programming model split between two architectures
- Cost of data movement between CPU and GPU will be reduced but still present
- Not scalable beyond a single chip





Research into scalable general purpose architectures



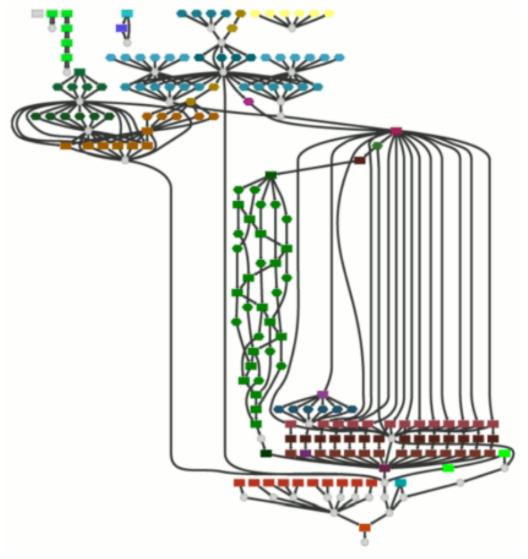


Simplify programming

Need general purpose parallel processors to simplify programming

Must support many algorithms, even within a single application e.g. Task (farms, pipeline) and data parallelism

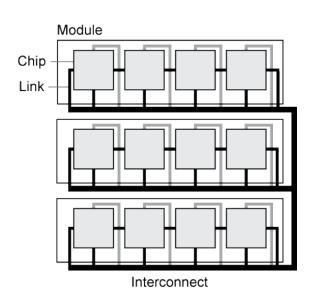
Performance must be comparable to special-purpose devices

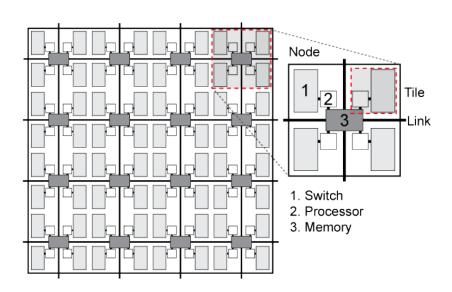




Performance must be scalable

- 1 1000 of cores per chip
 - Potentially millions of cores in a system
- Regular tiled implementation on chips, modules and boards







K Interconnect

- Must provide low latency, high throughput communication
- This must scale well with the number of processors
- Clos & hypercube networks provide these properties but it is assumed they are prohibitively difficult to build
 - Low dimensional meshes seem to be the convention
- Potential in new technology: 3D stacking, silicon substrates, optical interconnections, ...



& Summary

- Parallel languages are going through a renaissance
- Not just for the niche high-end any more
- No silver bullets, lots of "wheel reinventing"
- In HPC, GPUs being adopted quickly at the high-end
- In embedded computing, OpenCL gaining ground
- Movement towards high level general purpose models of parallelism



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