High Performance Computing with Accelerators

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Presentation Outline

• Recent trends in application accelerators
  • FPGAs, Cell, GPUs, …
  • Accelerator clusters

• Accelerator clusters at NCSA
  • Cluster management
  • Programmability Issues

• Production codes running on Accelerator Clusters
  • Cosmology, molecular dynamics, electronic structure, quantum chromodynamics
HPC on Special-purpose Processors

- **Field-Programmable Gate Arrays (FPGAs)**
  - Digital signal processing, embedded computing

- **Graphics Processing Units (GPUs)**
  - Desktop graphics accelerators

- **Physics Processing Units (PPUs)**
  - Desktop games accelerators

- **Sony/Toshiba/IBM Cell Broadband Engine**
  - Game console and digital content delivery systems

- **ClearSpeed accelerator**
  - Floating-point accelerator board for compute-intensive applications
GPU Performance Trends: FLOPS

- GeForce GTX 280
- Quad-core 3 GHz

Courtesy of NVIDIA
NVIDIA Tesla T10 GPU Architecture

- **T10 architecture**
  - 240 streaming processors arranged as 30 streaming multiprocessors
  - At 1.3 GHz this provides
    - 1 TFLOPS SP
    - 86.4 TFLOPS DP
  - 512-bit interface to off-chip GDDR3 memory
    - 102 GB/s bandwidth
NVIDIA Tesla S1070 GPU Computing Server

- 4 T10 GPUs
GPU Clusters at NCSA

- **Lincoln**
  - Production system available the standard NCSA/TeraGrid HPC allocation

- **AC**
  - Experimental system available for anybody who is interested in exploring GPU computing
Intel 64 Tesla Linux Cluster **Lincoln**

- **Dell PowerEdge 1955 server**
  - Intel 64 (Harpertown) 2.33 GHz dual socket quad core
  - 16 GB DDR2
  - InfiniBand SDR

- **S1070 1U GPU Computing Server**
  - 1.3 GHz Tesla T10 processors
  - 4x4 GB GDDR3 SDRAM

- **Cluster**
  - Servers: 192
    - CPU cores: 1536
  - Accelerator Units: 96
    - GPUs: 384
AMD Opteron Tesla Linux Cluster AC

- **HP xw9400 workstation**
  - 2216 AMD Opteron 2.4 GHz dual socket dual core
  - 8 GB DDR2
  - InfiniBand QDR
- **S1070 1U GPU Computing Server**
  - 1.3 GHz Tesla T10 processors
  - 4x4 GB GDDR3 SDRAM
- **Cluster**
  - Servers: 32
    - CPU cores: 128
  - Accelerator Units: 32
    - GPUs: 128
AC Cluster: 128 TFLOPS (SP)
NCSA GPU Cluster Management Software

- CUDA SDK Wrapper
  - Enables true GPU resources allocation by the scheduler
    - E.g., `qsub -l nodes=1:ppn=1` will allocate 1 CPU core and 1 GPU, fencing other GPUs in the node from accessing
  - Virtualizes GPU devices
  - Sets thread affinity to CPU core “nearest” the GPU device

- GPU Memory Scrubber
  - Cleans memory between runs

- GPU Memory Test utility
  - Tests memory for manufacturing defects
  - Tests memory for soft errors
  - Tests GPUs for entering an erroneous state
GPU Node Pre/Post Allocation Sequence

**Pre-Job (minimized for rapid device acquisition)**
- Assemble detected device file unless it exists
- Sanity check results
- Checkout requested GPU devices from that file
- Initialize CUDA wrapper shared memory segment with unique key for user (allows user to ssh to node outside of job environment and have same gpu devices visible)

**Post-Job**
- Use quick memtest run to verify healthy GPU state
- If bad state detected, mark node offline if other jobs present on node
- If no other jobs, reload kernel module to “heal” node (for CUDA 2.2 bug)
- Run memscrubber utility to clear gpu device memory
- Notify of any failure events with job details
- Terminate wrapper shared memory segment
- Check-in GPUs back to global file of detected devices
GPU Programming

- **3rd Party Programming tools**
  - CUDA C 2.2 SDK
  - OpenCL 1.2 SDK
  - PGI x64+GPU Fortran & C99 Compilers

- **IACAT’s on-going work**
  - CUDA-auto
  - CUDA-lite
  - GMAC
  - CUDA-tune
  - MCUDA/OpenMP
Implicitly parallel programming with data structure and function property annotations to enable auto parallelization

Locality annotation programming to eliminate need for explicit management of memory types and data transfers

Parameterized CUDA programming using auto-tuning and optimization space pruning

1st generation CUDA programming with explicit, hardwired thread organizations and explicit management of memory types and data transfers

Courtesy of Wen-mei Hwu, UIUC
GMAC – Designed to reduce accelerator use barrier

- **Unified CPU / GPU Address Space:**
  - Same CPU and GPU address

- **Customizable implicit data transfers:**
  - Transfer everything (safe mode)
  - Transfer dirty data before kernel execution
  - Transfer data as being produced (default)

- **Multi-process / Multi-thread support**

- **CUDA compatible**

*Courtesy of Wen-mei Hwu, UIUC*
GPU Cluster Applications

- **TPACF**
  - Cosmology code used to study how the matter is distributed in the Universe

- **NAMD**
  - Molecular Dynamics code used to run large-scale MD simulations

- **DSCF**
  - NSF CyberChem project; DSCF quantum chemistry code for energy calculations

- **WRF**
  - Weather modeling code

- **MILC**
  - Quantum chromodynamics code, work just started

- ...
TPACF on AC

QP GPU cluster scaling

<table>
<thead>
<tr>
<th>Quadro FX 5600</th>
<th>GeForce GTX 280 (SP)</th>
<th>Cell B./E. (SP)</th>
<th>GeForce GTX 280 (DP)</th>
<th>PowerXCell 8i (DP)</th>
<th>H101 (DP)</th>
</tr>
</thead>
<tbody>
<tr>
<td>88.2x</td>
<td>154.1x</td>
<td>59.1x</td>
<td>45.5x</td>
<td>30.9</td>
<td>6.8x</td>
</tr>
</tbody>
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speedup

execution time (sec)

number of MPI threads

ex. 567.5 289.0 147.5 79.9 46.3 29.3 23.7
NAMD on **Lincoln**
(8 cores and 2 GPUs per node, very early results)

![Graph showing performance comparison between CPU and GPU configurations.]

- **CPU (8ppn)**
- **CPU (4ppn)**
- **CPU (2ppn)**
- **GPU (4:1)**
- **GPU (2:1)**
- **GPU (1:1)**

- **2 GPUs = 24 cores**
- **4 GPUs**
- **8 GPUs**
- **16 GPUs**
- **8 GPUs = 96 CPU cores**

~5.6
~2.8

**CPU cores**

**STMV s/step**

**Courtesy of James Phillips, UIUC**
Bovine pancreatic trypsin inhibitor (BPTI)
3-21G, 875 atoms, 4893 basis functions

MPI timings and scalability

Courtesy of Ivan Ufimtsev, Stanford
GPU Clusters Open Issues / Future Work

- **GPU cluster configuration**
  - CPU/GPU ratio
  - Host/device memory ratio
  - Cluster interconnect requirements

- **Cluster management**
  - Resources allocation
  - Monitoring
  - CUDA/OpenCL SDK for HPC

- **Programming models for accelerator clusters**
  - CUDA/OpenCL+pthreads/MPI/OpenMP/Charm++/…