Introduction to GPU

Shin Kee Chung, Linqing Wen (UWA)
Kipp Cannon (Caltech)

What is GPU?

- GPU stands for Graphics Processing Unit
- Originally developed for graphics rendering
- Its development is mainly driven by the gaming market
- CUDA (Compute Unified Device Architecture) is the GPU programming language developed by NVIDIA
- AMD GPU was not tested due to the lack of library support (such as FFT library)

GPU and **CUDA**

- Popular
 - Over 100 million CUDA enabled GPU sold
- Easy to program using CUDA
 - C and C++ Integration
 - Sizeable computing libraries
 - CUDA Matlab Plugins
- Cost effective
 - \$400-\$500 can provide teraflops performance

CUDA Common Library

- CUFFT (CUDA FFT)
- CUBLAS (CUDA Basic Linear Algebra Subprograms)
- CUDA SDK (CUDA Software Development Kit)

http://www.nvidia.com/object/cuda learn.html

http://www.nvidia.com/object/cuda_sdks.html

Some CUDA SDK Examples

Monte Carlo Option Pricing

FFT Based 2D Convolution

Eigenvalues

Matrix Multiplications

N-body Simulation

CUDA in LAL - lalapps_inspiral

Developed by UWA-Caltech group

Shin Kee Chung, Linqing Wen (UWA), Kipp Cannon (Caltech)

• Results:

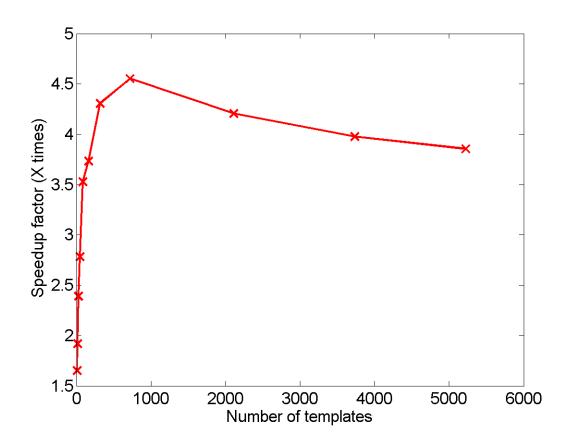
- Replace FFTW in lalapps_inspiral by CUDA FFT and demonstrated a 4x increase in speed overall using GeForce 8800 Ultra vs. 1 core of a 2.5 GHz Intel Core 2 Quad 9300 CPU.
- Implement chi-square statistic in lalapps_inspiral using direct calls to CUDA's multi-FFT routines and data parallelism feature
- Demonstrated a 16x increase in speed overall in lalapps_inspiral with the chi-square statistic turned on

• Publication:

Chung S. et al. 2009, CQG, to be published, arXiv:0906.4175 Chung, S. 2008, Honours thesis, UWA Chung, 2008 SURF report with Wen & Cannon

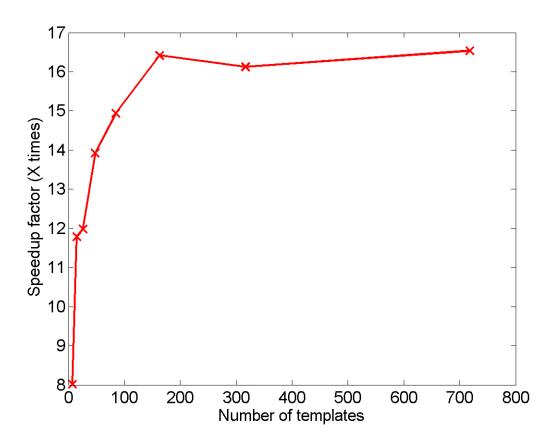
On going:

CUDA template parallel lalapps_inspiral



lalapps_inspiral Speedup With CUFFT

We achieve more than 4 times speedup simply by applying CUFFT in LAL.



lalapps_inspiral Speedup With CUDA Chi-square Implementation

We modified the original Chi-square in lalapps_inspiral to utilize the multi-FFT routines and data parallelism feature of CUDA.

We achieved more than 16 times speedup.

CUDA in LAL — Current Status

Enable FFT package's CUDA back-end with

```
$ ./configure --with-cuda={path to cuda} ...
```

replaces all single precision FFTs in LAL/LALApps codes with GPU-accelerated versions.

- Works on
 - Linux machines with CUDA installed.
 - Macs machines with CUDA installed.
- CUDA-based LAL/LALApps tested with
 - GeForce 8500 GT (card purchased by D. Brown for testing)
 - GeForce 8800 Ultra (Shinkee's development box at CIT)
 - GeForce 8800 GTX, GTX 285 (UWA)
 - GeForce 8600M GT on Mac OS X Leopard laptop
 - GTX 295 (RMKI, Hungary)
- "Test" = obtain correct output from lalapps_inspiral (fractional error < 0.03% for Shinkee's tests).

GPU Low-Latency Inspiral Search

- UWA-Caltech collaboration
- Applied to time-domain low-latency infinite impulse response (IIR) filter
 - Hooper et al. 2009, Budapest LSC Meeting LIGO-G0900770-v2 https://dcc.ligo.org/cgi-bin/private/DocDB/ShowDocument?docid=5333

Results:

- 5 times speedup is achieved
- Tested using GTX 285 and 2.4 GHz Intel Core 2 Quads 6600 CPU (one core)

Future Work:

- Apply to stream-based time-domain low-latency pipeline LLOID developed by Caltech LIGO group
 - Insert GPU-enabled IIR filter
 - Replace individual elements in LLOID with GPU-enable components

- LSC GPU Discussion Group
 - Mailing List
 - GPU-Discuss@ligo.gwastro.psu.edu
 - Still a starting phase in LSC
 - ~30 subscribers, ~15 LSC institutions
 - » UWA, MPI/AEI, Northwestern, Caltech, LLO, RMKI Virgo Group (Hungary), Tsinghua U. (China), PSU, UWM, UNH, RIT, Umass, Cardiff, Birmingham, Columbia, ANU, U. Michigan
 - Wiki page:
 - https://www.lscgroup.phys.uwm.edu/daswg/wiki/GPUDevelopment

- U. of Western Australia (UWA)
 - Linqing Wen, Shin Kee Chung*, Shaun Hooper, David Blair, Amitava Datta
 - GPU-enabled inspiral search pipeline (previous slides)
 - Resource
 - Tested on Geforce 8800, FX 1700, GTX 285, GTX 295 (on 1 card)
 - International Center for Radio Astronomy Research (ICRAR), WASP center at UWA, CSIRO GPU cluster (200 GPUs)

http://www.gravity.uwa.edu.au/gpu_implementation.html

Caltech

- Kipp Cannon,
- GPU-enabled inspiral search pipeline and LLOID (previous slides)
- Support from the LIGO group

AEI-Hannover: Einstein @ Home

Bruce Allen, Reinhard Prix, Oliver Bock, Bernd Machenschalk,
 Carsten Aulbert et al.

• Hardware:

- 6 machines with two Tesla C1060 each (3 Intel, 3 AMD, all Debian Lenny x64)
- 1 machine with two GTX 285 (AMD, running XP32, Vista32 and Debian x64)
- 1 machine with one dual-GPU GTX 295 (Intel, running XP64, Vista64 and Debian x64)
- The ATLAS cluster is going to be extended by ~120 Tesla cards later this year or early next year (available to LSC members, same resource policy as for ATLAS itself)

GPU development

- hierarchical search for continuous GWs in S5 data (development)
- the search for binary pulsars in Arecibo radio data (improvement, public beta).

Northwestern

- Vicky Kalogera et al.
- 30-GPU cluster: "happy to make them available"
- benchmarking the SPINspiral MCMC code

RMKI Virgo Group (Hungary)

- Debreczeni Gergely et al.
- Testing on NVIDIA GeForce GTX 295.
- Submitted proposals for GPU clusters to OTKA ("Hungarian NSF")

University of New Hampshire / NIKHEF

- Maurik Holtrop, Jo van den Brand
- Hardware: two GTX280 and one 8800 GT, Tesla cluster (future)
- Application: CW all sky search using quadratic filters.

- Tsinghua U. (China)
 - Junwei Cao et al.
 - Hardware
 - GeForce 9800GTX
 - Substantial non-LSC related GPU research has been done
 - implemented a collection of algorithms using CUDA
 - Confirm that CUDA is a good platform for computation-intense applications while performs not very well for control-intense algorithms
 - investigate different GPU programming models
 - Confirm that only some GPU programming models are good extension of CUDA
 - designed and implemented a parallel Viterbi sequence finding algorithm on GPU
 - More than one order magnitude is achieved
 - optimization GPU for finance application
 - 100 fold speedup achieved
 - MapReduce framework (MARS) on GPU
 - Currently planning to apply GPU to burst search using the Omega pipeline

UWM

- Patrick Brady, Adam Mercer et al
- Adam helps build system patches, committed LAL CUDA FFT and working on committing GPU-enabled chi-square test
- plan to deploy a substantial GPU testbed at UWM over the next 6-12 months to enable larger scale prototyping of GPU enabled codes for gravitational-wave astronomy.

LLO

- Dwayne Giardina, Rupal Amin et al
- built a workstation at LLO with a NVIDIA Tesla C1060
- Compare a Matlab scripts with and without CUDA plugins

Acknowledgements

- Alan Weinstein, Patrick Brady and Sam Finn for support
- Inputs from Adam Mercer, Oliver Bock, Jun-wei Cao, Debreczni Gergely, Vivien Raymond, Reinhard Prix, and Maurik Holtrop
- GPU-Discuss community

Appendix

GPU vs CPU

GPU

- Its architecture adopts parallel programming naturally
- Less efficient in flow control
- Small cache size limits the speed of memory access

CPU

- Originally designed to execute commands sequentially
- Very efficient in flow control
- Huge cache size speed up its memory access



Comparison of CPU and GPU hardware

Diagram taken from the NVIDIA CUDA Programming Guide.

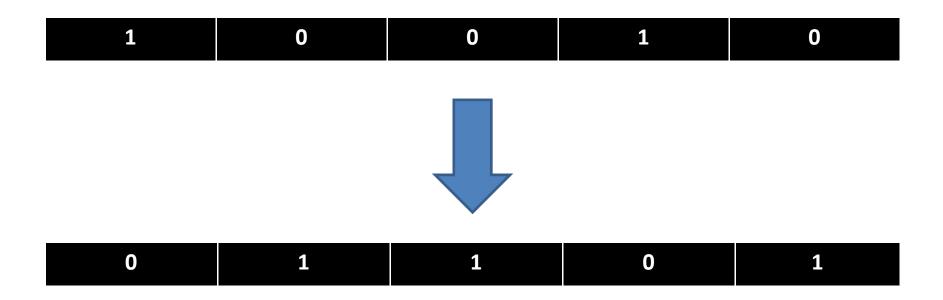
Graphics Cards Specification

GPU Card	Peak Performance	Memory Size	Memory Bandwidth	Number of Processing Cores	Max Power Consumption	Price (USD)
GTX 285 (Single GPU)	1062.72 GFLOPs	1 GB	159.0 GB/sec	240	204W	~ \$400
GTX 295 (Dual GPU)	1788.48 GFLOPs	1792 MB (896 MB each)	223.8 GB/sec	480 (240 each)	289W	~ \$550
Supercomputer with up to 4 Tesla C1060	Up to 4 TFLOPs	4 GB each	102 GB/sec each	240 each	187.8W each	~ \$8000

CUDA Quick Start Guide

- After installing CUDA following some simple instructions from NVIDIA website:
 - Write a simple "hello world" C program, and save it as .cu extension
 - Compile it using nvcc
 nvcc -o program_name program_name.cu
 - This is already a CUDA program (although it is still executed in CPU)
 - Then we can start putting in kernel functions (that runs in GPU) s

Example: Simple Number Inverter



Example: Simple Number Inverter

```
C Program
int *run(int *data, int length)
    int i;
    for( i = 0; i < length; i++)
                                            Calculation
                                            done in a loop
        data[i] = 1 - data[i];
    return data;
```

Example: Simple Number Inverter

CUDA Program, multi-threaded execution Kernel Function global___ void invert(int *d_data, int length) // Getting the thread id, block id and number of threads per block int tx = threadIdx.x; int bx = blockIdx.xiint numThreads = blockDim.x; No loop is used Inverting the element accessed by each thread d_data[bx * numThreads + tx] = 1 - d_data[bx * numThreads + tx]; **Host Function** int *run(int *data, int length) // Perform inversion, assuming that total threads = array length invert <<< blocks, threads >>>(d data, length); // Then copy d_data to data return data;

GPU in Science: example

N-body Simulation

- Pre-CUDA:
 - Portegies Zwart et al. 2007New Astron., 12, 641
- CUDA:
 - Belleman et al. 2008 New Astron., 13, 103
- GeForce 8800GTX with CUDA runs at about the same speed as GRAPE-6Af for N ≥ 10⁶

Radio Astronomy

- GPU enabled correlator
 - Harris et al. 2008
 Experimental Astron., 22, 129
- > 100x achieved
- International Center for Radio Astronomy Research (ICRAR), UWA and Curtin

Possible Future Application

- lalapps inspiral is unable to scale up to Adv.LIGO analysis requirements; a re-work of the internal data management is required.
- One project to investigate solutions is "gstlal", a project to combine LAL with GStreamer, a Free stream-based multimedia signal processing framework.
- Prototype application is LLOID (a Low-Latency Online Inspiral Data analysis application) - a stream-based version of lalapps inspiral.
 - allows for very long, low-frequency, templates
 - provides sub-template latency
 - allows matched-filtering across gaps in data
 - multi-threaded to take advantage of multi-core CPUs

Possible Future Application

- gstreamer/gstlal provides collection of "elements" that are chained together in a graph to construct the analysis pipeline.
 - highly-modular
 - example elements: resampler, tee, adder, FIR filter, IIR filter, mixer.
 - individual elements are easily replaced with alternate implementations e.g.,
 GPU-based versions.

GPU use speculative:

- thread contention for GPU is potentially a problem.
- data rate on PCI bus is potentially a problem.
- increasing on-device processing addresses PCI bandwidth. Possible?
- gstreamer supports concept of special on-device data buffers; possibly allows smart management of GPU RAM.
- or maybe data rate on PCI bus will be fine.
- many unknowns.