Running Geant4 Functions on a GPU

Discussion of Results

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Overview

- 1 Introduction
 - Brief Project Overview
 - Explanation of Terms
 - Scope
 - Purpose
- 2 Features
 - Easily Enable/Disable GPU Acceleration
 - Impl. 1: Existing Module in GPU Memory
 - Impl. 2: Add New GPU-Accelerated Methods to Interface
 - Accuracy / Testing
- 3 Conclusion
 - Summary of Results
 - Recommendations

Brief Project Overview

Take an existing particle simulation toolkit - Geant4 - and have some functions run on a GPU device to improve performance.

What is Geant4?

- Geant4 is a toolkit that is meant to simulate the passage of particles through matter.
- It has been developed over the years through collaborative effort of many different institutions and individuals.
- Geant4's diverse particle simulation library has a wide variety of applications including
 - High energy physics simulations
 - Space and radiation simulations
 - Medical physics simulations

Demonstration

Demonstration – Running Geant4 on the CPU Hadr04 With Visualization

What is GP-GPU Computing?

- General-purpose graphic-processing-unit computing is a re-purposing of graphics hardware
- Allows GPUs to perform computations that would typically be computed on the CPU
- If a particular problem is well suited to parallelization,
 GP-GPU computing can greatly increase performance

Scope

- Make current CPU functions available for use on GPU
 - Add appropriate prefixes to function definitions
 - Make use of multiple parallel threads to execute each function
- Ensure correctness of each GPU available function by matching results to the corresponding CPU function
- Compare performance of GPU available functions to CPU functions

Purpose

- Determine if target functions are suitable to parallelization
- Increase performance of functions when run on GPU
- Decrease time required to run simulations involving ported functions

casniy Enable/Disable GPU Acceleration mpl. 1: Existing Module in GPU Memory mpl. 2: Add New GPU-Accelerated Methods to Interfac Accuracy / Testing

Features

- GPU acceleration available on an "opt-in" basis
- Easy to enable/disable GPU acceleration
- If GPU acceleration is enabled, some methods will run on GPU
- Same results whether acceleration enabled or disabled

Easily Enable/Disable GPU Acceleration

- Existing projects can use GPU acceleration without having to change any code
- Flag during build phase enables/disables GPU acceleration
- No new functions to learn ¹

¹implementation 1 only

Easily Enable/Disable GPU Acceleration
Impl. 1: Existing Module in GPU Memory
Impl. 2: Add New GPU-Accelerated Methods to Interface
Accuracy / Testing

Demonstration

Demonstration - Enabling CUDA Acceleration

Easily Enable/Disable GPU Acceleration

Method calls to G4ParticleHPVector forwarded to GPU-based implementation

■ This decision is made at compile time based on cmake flag

Example of Forwarding Method Calls

```
inline G4double GetY(G4double x)
{
    #if GEANT4_ENABLE_CUDA
        return cudaVector->GetXsec(x);
    #else
        return GetXsec(x);
    #endif
}
```

Easily Enable/Disable GPU Acceleration
Impl. 1: Existing Module in GPU Memory
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Accelerating Module on GPU

Existing module G4ParticleHPVector ported to GPU using CUDA

Definition: CUDA

CUDA is a GP-GPU programming model developed by NVIDIA, for use with NVIDIA graphics cards

Why G4ParticleHPVector?

- Represents empirically-found probabilities of collisions for different particles based on their energy
- Identified as starting point by relevant stakeholders
 - Used heavily in simulations run by stakeholders
- Seems well-suited to parallelization
 - Based on large vector of 2D points
 - Performs calculations over this vector
 - Sorted by x-value (particle energy)

Two Implementations

- Forward all calls to existing G4ParticleHPVector interface to a GPU-based implementation of the module
 - Store data vector in GPU memory
 - Copy results back to the CPU to return to the caller
- 2 Add new methods to G4ParticleHPVector interface that are well-suited to GPU computing
 - Copy data vector to GPU memory on method call
 - Existing G4ParticleHPVector methods unchanged, continue to run on CPU

Easily Enable/Disable GPU Acceleration
Impl. 1: Existing Module in GPU Memory
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Accuracy / Testing

Impl. 1: Existing Module in GPU Memory

Calls to G4ParticleHPVector forwarded to new GPU-based class

Pros:

- + Do not have to maintain a copy of the vector on the CPU
- + Do not have to maintain a hashed vector
- + Reduces how much is being copied to the GPU

Cons:

All methods are run on the GPU

Easily Enable/Disable GPO Acceleration

Impl. 1: Existing Module in GPU Memory

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Accuracy / Testing

Demonstration

Hadr04 with Visualization

Easily Enable/Disable GPU Acceleration

Impl. 1: Existing Module in GPU Memory

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Accuracy / Testing

Impl. 1 - Times

Times_CUDA

```
int tid = blockDim.x * blockIdx.x + threadIdx.x;
if (tid < nEntries)
    theData[tid].xSec = theData[tid].xSec * factor;</pre>
```

Easily Enable/Disable GPU Acceleration

Impl. 1: Existing Module in GPU Memory

Impl. 2: Add New GPU-Accelerated Methods to Interface

Accuracy / Testing

Impl. 1 - GetXSec

GetXSec_CUDA

```
int start = (blockDim.x * blockIdx.x + threadIdx.x);
for (int i = start; i < nEntries; i += numThreads)
  if (theData[i].energy >= e) {
    resultIndex = Min(resultIndex, i);
    return;
  }
```

Impl. 1: Performance Results Summary

- Most methods slower on GPU until ~10,000 entries in data vector
- Most commonly-used methods significantly slower on GPU, even with large data vector
 - Lots of data accesses
- Many problems in vector class not well-suited to parallelism

Impl. 1: Performance Results - Times

Multiplies each point in vector by factor

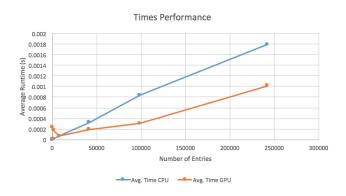


Figure: Runtime vs. Number of Data Points - Times

Impl. 1: Performance Results - GetXSec



Figure: Runtime vs. Number of Data Points - GetXSec

Impl. 1: Performance Results - SampleLin

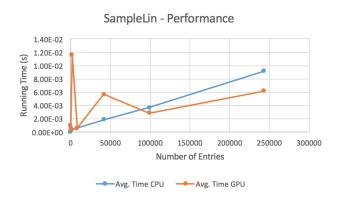


Figure: Runtime vs. Number of Data Points - SampleLin

Easily Enable/Disable GPU Acceleration
Impl. 1: Existing Module in GPU Memory
Impl. 2: Add New GPU-Accelerated Methods to Interface
Accuracy / Testing

Impl. 1: Performance Results – System Tests

System Test #1:

Table: Performance - Water, 2000 events

CPU Time GPU Time Speedup of GPU 54.55s 72.08s -1.32×

System Test #2:

Table: Performance - Uranium, 2000 events

Impl. 1: Performance Results – System Tests (Cont.)

System Test #3:

Table: Performance - Water, 2000 events

CPU Time GPU Time Speedup of GPU 54.55s 72.08s -1.32×

System Test #4:

Table: Performance - Uranium, 2000 events

Easily Enable/Disable GPU Acceleration
Impl. 1: Existing Module in GPU Memory
Impl. 2: Add New GPU-Accelerated Methods to Interface
Accuracy / Testing

Impl. 2: Performance Results – System Tests (Cont.)

System Test #5:

Table: Performance - Water, 2000 events

CPU Time GPU Time Speedup of GPU 54.55s 72.08s -1.32×

System Test #6:

Table: Performance - Uranium, 2000 events

Easily Enable/Disable GPU Acceleration

Impl. 1: Existing Module in GPU Memory

Impl. 2: Add New GPU-Accelerated Methods to Interfac

Accuracy / Testing

Impl. 1: Performance Discussion

■ Simple "getters" and "setters" now

Easily Enable/Disable GPU Acceleration
Impl. 1: Existing Module in GPU Memory
Impl. 2: Add New GPU-Accelerated Methods to Interface
Accuracy / Testing

Impl. 2: Add New GPU-Accelerated Methods to Interface

Add new methods to G4ParticleHPVector interface that are well-suited to parallelism

Pros:

- + Only methods that run faster on the GPU are implemented
- + Not forced to run methods that run slowly on GPU

Cons:

- Will have to maintain two copies of the vector
- More copying the vector to and from the GPU

Impl. 2: GetXSecList

- Fill an array of energies for which we want to cross section values
- Send the array to the GPU to work on
- Each thread work on its own query(s)

Implementation - GetXSecList

GetXSecList

```
stepSize = sqrt(nEntries);
i = 0;
e = queryList[threadID];

for (i = 0; i < nEntries; i += stepSize)
   if (d_theData[i].energy >= e)
        break;
```

Implementation - GetXSecList -- cont

```
GetXSecList - cont
i = i - (stepSize - 1);
for (; i < nEntries; i++)
   if (d_theData[i].energy >= e)
        break;
d_queryList[threadID] = i;
```

Impl. 2: Performance Results Summary

Easily Enable/Disable GPU Acceleration Impl. 1: Existing Module in GPU Memory Impl. 2: Add New GPU-Accelerated Methods to Interface Accuracy / Testing

Impl. 2: Performance Results - GetXSecList

Impl. 2: Performance Results – System Tests

System Test #1:

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Impl. 2: Performance Results – System Tests (Cont.)

System Test #3:

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Impl. 2: Performance Results – System Tests (Cont.)

System Test #5:

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System Test #6:

Table: Performance - Uranium, 2000 events

Impl. 2: Performance Discussion

Accuracy

- All modified functions except SampleLin and Sample yield results that precisely match
- Some functions fell extremely close in accuracy to the original, and were considered to 'pass'
- The average of 1000 SampleLin tests deviated from the average of 1000 tests of the original with an error of 0.01
- The system tests differ if the number of nentries is greater than 500; if not however the results of the system test conform.

Accuracy

- The deviations in SampleLin and Sample can be attributed to the functions use of random numbers
- The negligable deviations in other ported functions are likely attributed to differences in CPU and GPU arithmetic, leading to different round-off errors

Testing

- Comparing test results and performance with GPU acceleration enabled and disabled
- Testing framework based on two phases, one program for each phase
 - GenerateTestResults: Run unit tests and save results to file
 - 2 AnalyzeTestResults: Compare results from CPU and GPU
- Run GenerateTestResults once for GPU acceleration enabled, once with it disabled

Easily Enable/Disable GPU Acceleration
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GenerateTestResults Details

- Includes testing version number in results file for analysis stage
- Outputs simple results directly to results file
- For vectors, calculates hash for vector and output it
- Outputs timing data to separate file

Example: Snippet of Generated Test Results

```
#void G4ParticleHPVector_CUDA::GetXsecBuffer(
   G4double * queryList, G4int length)_6
@numQueries=10
hash: 16548307878283220284
@numQueries=50
hash: 3204132713354913775
```

Demonstration

Demonstration - Generating Test Results

AnalyzeTestResults Details

Two main functions:

- Compare results for each test case, printing status to stdout
 - If test failed, output differing values
 - Summarize test results at the end with number passed
- Generate .csv file from timing data
 - One row per unique method call, columns show CPU time,
 GPU time, method name and parameters
 - Can use Excel to analyze performance results

Demonstration

Demonstration - Analyzing Test Results

Summary of Results

Recommendations