

# Running Geant4 Functions on a GPU

## Discussion of Results

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# Overview

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# 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

# What is GP-GPU

- 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

# Features

- GPU acceleration available on an “opt-in” basis
- Easy to enable/disable GPU acceleration
- 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 <sup>1</sup>

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<sup>1</sup>implementation 1 only

# Implementation

- Header forwards function calls to GPU or CPU Implementation
- This decision is made at compile time.

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

# Accelerating Module on GPU



# 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

- 1 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

## 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

# Implementation – Times

# Implementation – GetXSec



# Implementation – SampleLin

# 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

## Performance Results – Times

- Multiplies each point in vector by factor

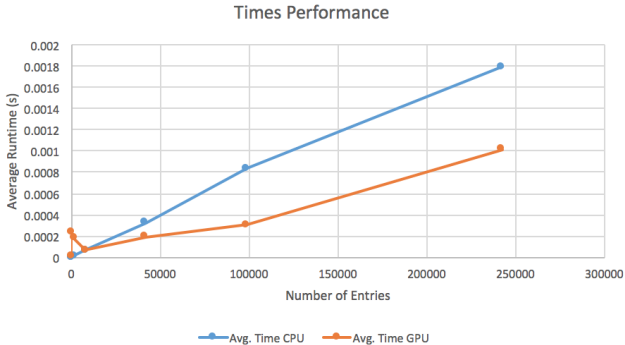


Figure: Runtime vs. Number of Data Points – Times

# Performance Results – GetXSec

# Performance Results – SampleLin

# Performance Results – System Tests

# Performance Discussion

## 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



## Implementation – GetXSecList

- Fill an array of energies we want to get xSec values for
- Send the array to the GPU to work on

# Performance Results Summary

# Performance Results – GetXSecList

# Performance Results – System Tests

# Performance Discussion

# Accuracy

# Testing

# Summary of Results



# Recommendations