

# Geant4 based simulation of radiotherapy in CUDA

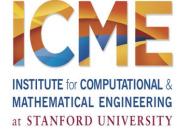
#### Koichi Murakami (KEK / CRC)

#### Stanford ICME, SLAC, G4-Japan Collaboration supported by NVIDIA

#### The Collaboration

## Geant4 @ SLAC





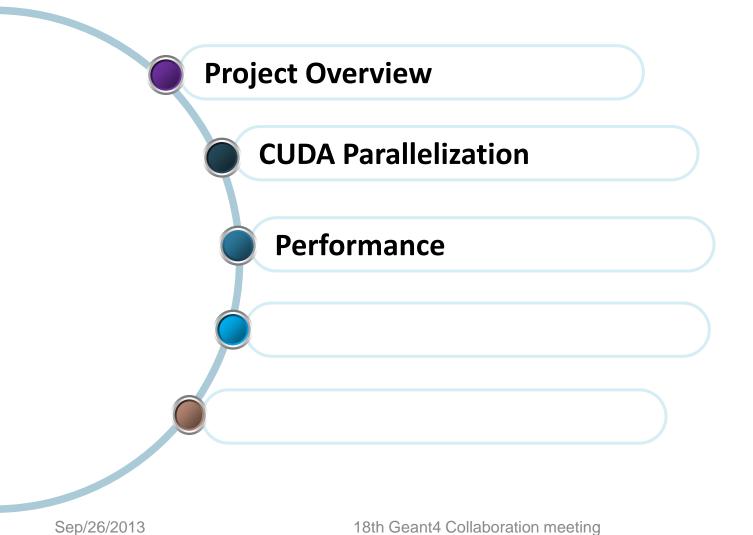




Special thanks to the CUDA Center of Excellence Program

18th Geant4 Collaboration meeting

#### Contents

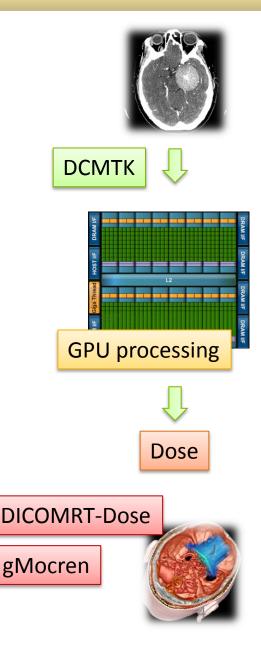


3

## G4CU Project Overview

#### **Dose calculation for radiation therapy**

- GPU-powered
  - parallel processing with CUDA
  - boost-up calculation speed
  - CUDA porting of Geant4
- Functions
  - voxel geometry
    - including DICOM interface
    - material : water with variable densities
  - limited Geant4 EM physic processes
    - electron/positron/gamma
  - scoring dose in each voxel



## **Physics Processes**

- particles : electron, positron, gamma
- energy range up to 100 MeV
- material: water with variable densities
- processes:

#### electron / positron

- energy loss (ionization, bremsstrahlung)
- multiple scattering
- positron annihilation

#### gamma

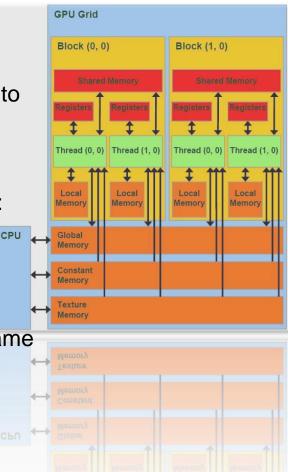
- Compton scattering
- photo electric effect
- gamma conversion

#### physics tables

- dE/dx, range, etc are retrieved from Geant4
- Physics tables are prepared for "standard" water and rescaled with the density of each voxel.

## **CUDA Basics**

- "SIMD" architecture : Single Instruction, Multiple Data
  - CUDA is a data parallel language
  - wants to run same instruction on multiple pieces of data
- Coalesced memory access
  - to maximize memory throughput, we want a single read to satisfy as many threads as possible
- Memory hierarchy
  - CUDA provides access to several device memory types:
    - global, shared, constant, texture
  - better memory usage for better performance
- Race conditions
  - arise when multiple CUDA threads attempt to write to same location in global memory
  - may happen in dose accumulation
  - we avoid race conditions or using atomic operations



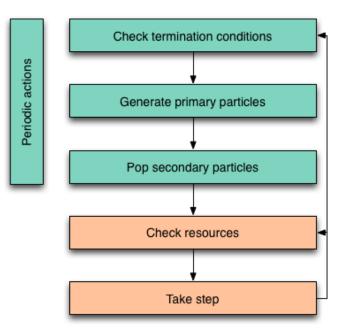
6

#### Parallelization Challenges in Geant4

- Programming Methodology
  - CUDA porting of large and complex code
    - large scale of object-oriented design
    - inter-dependencies in many places
  - Branching, look-up tables, single-thread optimizations
- Software Complexity
  - Sophisticated geometry and tracking management
  - Elaborate physics models

#### **G4CU** Basics

- Each GPU thread processes a single track until it dies or exits
  - GPU runs on 32k CUDA threads (256 threads on 128 bloks)
  - A track stores data for
    - particle spices, position, direction, energy, etc
- Each thread has two stacks
  - one for storing secondary particles
  - one for recording the energy dose in a voxel
- Periodic actions:
  - check termination conditions
  - generate primaries
  - pop secondary particles



#### Notes on Dose Accumulation

- 2 critical issues on performance:
- Race conditions might arise when multiple CUDA threads attempt to write to same location in global memory.
  - That may happen in dose accumulation in each voxel.
  - Two ideas were tried:
    - parallel stack for dose and reduction
    - atomicAdd operation in CUDA -> better performance
      - explicit memory access
- *Double precision* variables are used for dose.
  - other variables are single precision.
  - prevent from overflow
    - small energy dep. + large accumulated dose

## Performance Profiling

#### nvpof is helpful to identify performance bottlenecks.

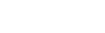
Process	Process total	Post step	PIL	Along step	Manag.	Init.	Step length	At-rest
Component total (%)	100.00	52.80	20.73	14.16	4.19	3.78	3.46	0.89
Bremsstrahlung	32.81	29.83	2.23			0.75		
Ionization	14.83	1.70	3.50	8.84		0.79		
Photo-electric effect	10.79	8.80	1.57			0.41		
Gamma conversion	10.67	8.72	1.54			0.41		
Multiple scattering	10.50		3.43	4.58			2.49	
Transport	8.67	1.17	5.23	0.74		0.57	0.96	
Compton scattering	4.20	2.14	1.58			0.48		
Management	4.19				4.19			
Pair production	2.56	0.44	1.23			0.36		0.53
Electron deletion	0.79		0.43					0.36

#### Hardware & SDK

- GPU:
  - Tesla K20 (Kepler)
  - 2496 cores, 706 MHz, 5GB GDDR5 (ECC)
- SDK:
   CUDA 5.5, CURAND
- CPU:

Sep/26/2013

- Xeon X5680 (3.33GHz)



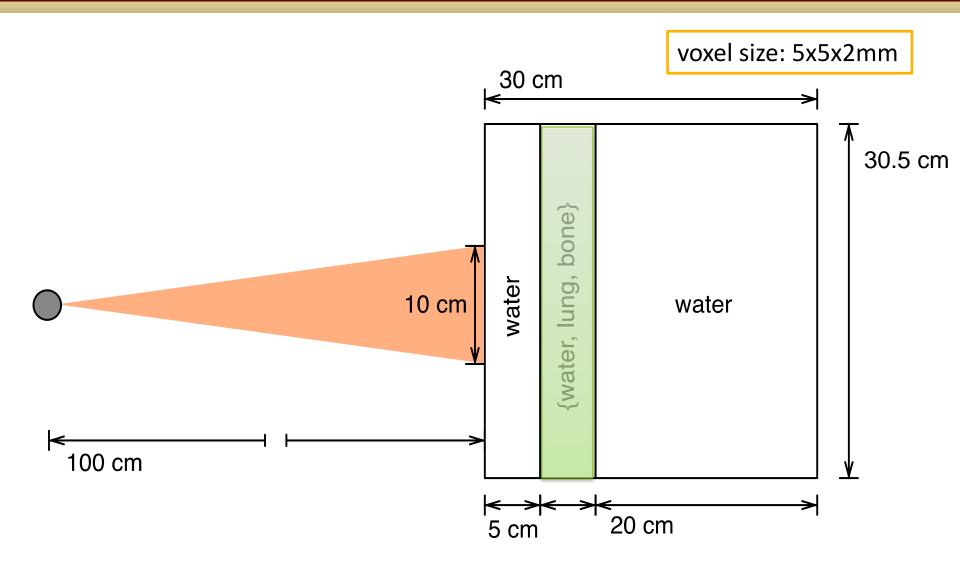






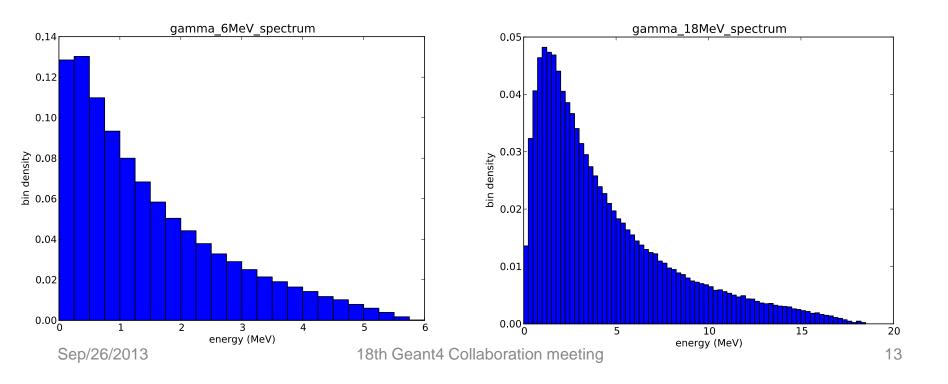


#### Phantom Geometry Configurations



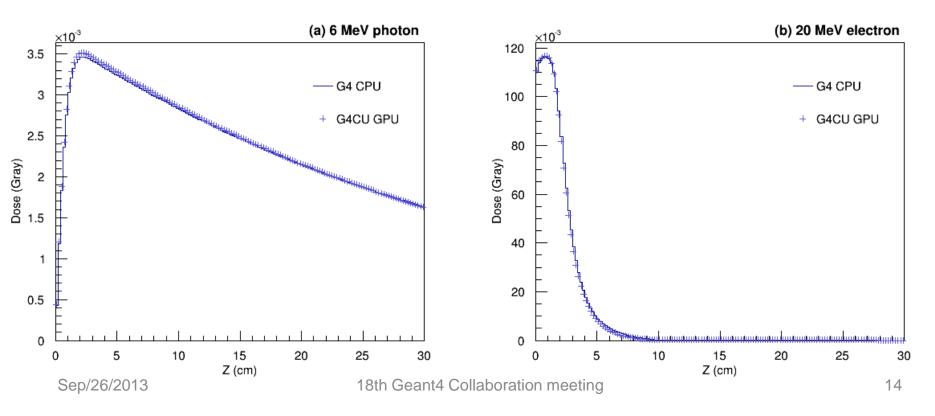
#### **Particle Sources**

- Mono-energetic source :
  - 20 MeV electron
  - 6 MeV photons
- Spread energy source generated by medical linac
  - 6 MV & 18 MV photons

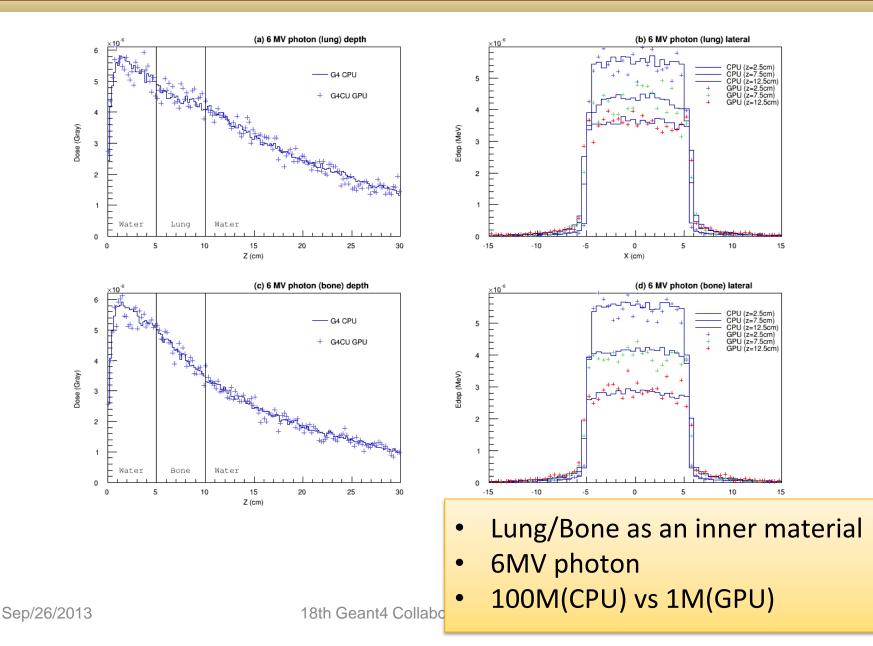


#### Depth Dose Distribution, CPU vs. GPU

- Depth dose distribution for water phantom
  - 6MeV photon and 20 MeV electron (pencil beam)
  - depth dose along central voxels
  - 100M primaries



#### **Comparisons for Slab Phantoms**

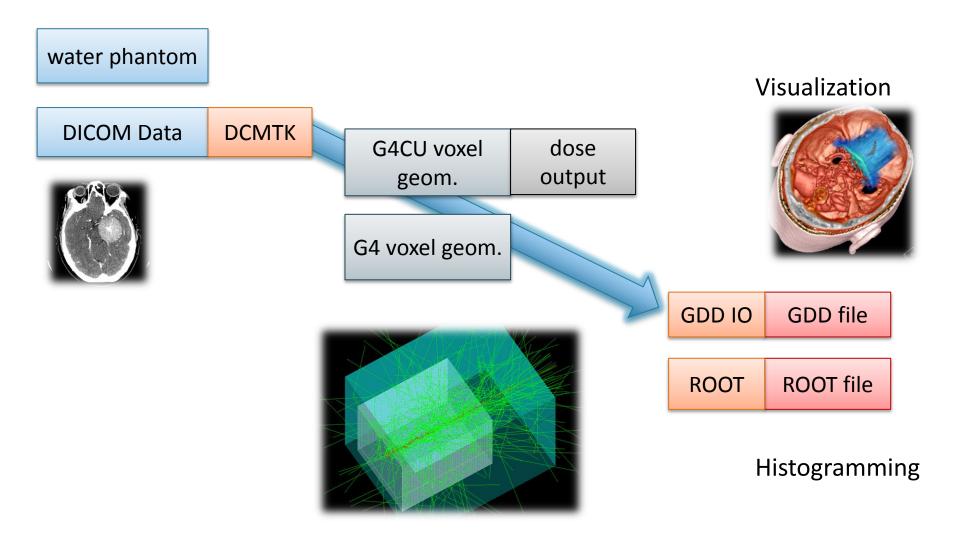


#### Computation Time of Geant4/CPU and G4CU/GPU

Primary	Phantom	Time/History CPU (sec)	Time/History GPU (sec)	CPU/GPU
20 MeV electron	Water	1.06E-03	2.52E-05	42.1
6 MeV photon	Water	4.47E-04	1.12E-05	39.9

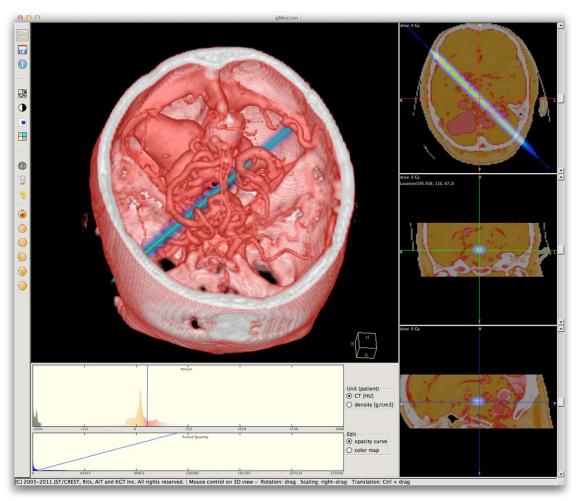
CPU: Intel Xeon X5680 (3.33 GHz) GPU : NVIDIA Tesla K20

## G4CU achieves about *40 times speedup* against CPU-based Geant4 simulation.



## Visualization with gMocren

#### 50M 6MeV photons, pencil beam shot on head



voxel size: 1.7 x 1.7 x 1.2 mm segments: 128 x 128 x 64 ≈ 1M voxels

Just for Demonstration

## Summary

- Collaborative activity on Geant4-GPU between

   Stanford ICME, SLAC, and G4-Japan (KEK), supported by NVIDIA
- Focused on medical application
  - Dose calculation in voxel domain
  - Geant4 EM physics processes
- CUDA porting of Geant4
  - Core part of implementation was done.
  - Verification against CPU version of Geant4
    - well-agreed in 1<sup>st</sup> order
  - Performance gain about 40 times
- Several ongoing follow-ups
  - robustness, performance improvements, improved code