





AdePT

Accelerated demonstrator of electromagnetic Particle Transport

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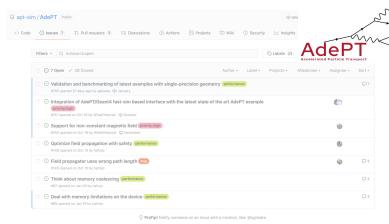
Targets



- Functionality: make all simulation components work on GPU
 - Physics, geometry, field, but also scoring code to limited extent
 - Prototype e⁺, e⁻ and γ EM shower simulation on GPU
- Correctness: validate results and ensure reproducibility
 - Against Geant4 equivalent
- Usability: integrate in a hybrid CPU-GPU Geant4 workflow
 - For realistic experimental setups
- Performance: understand/address bottlenecks limiting performance
 - Estimate feasibility and effort for efficient GPU simulation

The project

- GitHub <u>repository</u>
 - Initial commit in Sep 2020, $\mathcal{O}(10)$ contributors
- Strategy: integrate gradually features as new examples
 - No library build, maximize flexibility to explore different directions
- Few external dependencies
 - Geometry: <u>VecGeom</u> library, enhancing GPU-related features
 - Physics: <u>G4HepEm</u> library, a GPU-friendly port of Geant4 EM interactions
- Portability aspects not a major priority in this project phase
 - Preliminary investigations started with Alpaka and OneAPI



The prototype



Initial commit		Sep 2020
		Oct 2020
Fisher-Price like example & Alpaka version of it		
Simple workflow with geometry navigation (exa2)	-	_ Dec 2020
First example working in constant field (exa4)		_ Feb 2021
First integration with G4HepEm physics (exa5)		
First example with geometry and Bz field (exa6)		
Added gamma interactions (exa9), added TestEm3 similar to Geant4		_ Mar 2021
Added import/export of geometry and physics between Geant4 and AdePT (exa7)	-	_ Jun 2021
MT version of TestEm3		_ Jul 2021
Single-precision support in geometry		_ Sep 2021
Standalone example with generalized GDML geometry (exal3)		_ Oct 2021
Added support for multiple scattering		
Integration with Geant4 workflow demonstrator (exa14)		_ Feb 2022
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GPU-friendly rewrite of EM physics



- ► G4HepEm: compact library of EM processes for HEP
 - Covers the complete physics for e^- , e^+ and γ particle transport
 - Initialization of physics tables dependent on Geant4, but usage on GPU standalone and lightweight
- Design of library very supportive for heterogeneous simulations
 - Interfaces: standalone functions without global state
 - Data: physics tables and other data structures copied to GPUs
 - Reusing > 95% of the code from G4HepEm for GPU shower simulation

Interactions modelled for e⁺, e⁻ and y

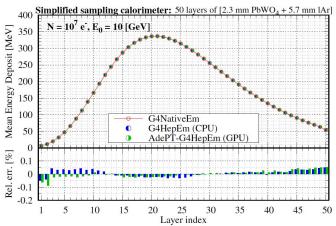
Particle	Interactions	Models	Geant4 (EM-Opt0)	G4HepEm (with G4HepEm prefix)	Energy Range
e-	Ionisation	Moller	G4MollerBhabhaModel	ElectronInteractionIoni	1 keV - 100 TeV
	Bremsstrahlung	Seltzer-Berger	G4SeltzerBergerModel	FlactronInteractionProm	1 keV - 1 GeV
		Rel. model ¹	G4eBremsstrahlungRelModel	ElectronInteractionBrem (including both models)	1 GeV - 100 TeV
	Coulomb scat.	Urban	G4UrbanMscModel	ElectronInteractionUMSC	1 keV - 100 MeV
		Wentzel-VI	G4WentzelVIModel		100 MeV - 100 TeV
e^+	Ionisation	Bhabha	G4MollerBhabhaModel	ElectronInteractionIoni	1 keV - 100 TeV
	Bremsstrahlung	Seltzer-Berger	G4SeltzerBergerModel	FlactronInteractionProm	1 keV - 1 GeV
		Rel. model	G4eBremsstrahlungRelModel	ElectronInteractionBrem (including both models)	1 GeV - 100 TeV
	Coulomb scat.	Urban	G4UrbanMscModel	ElectronInteractionUMSC	1 keV - 100 MeV
		Wentzel-VI	G4WentzelVIModel		100 MeV - 100 TeV
	Annihilation	$e^+ - e^- \rightarrow 2\gamma$	G4eplusAnnihilation	PositronInteractionAnnihilation	0^3 - 100 TeV
γ	Photoelectric	Livermore	G4LivermorePhotoElectricModel	GammaInteractionPhotoelectric ⁴	0 ⁵ - 100 TeV
	Compton scat.	Klein - Nishina ⁶	G4KleinNishinaCompton	GammaInteractionCompton	100 eV - 100 TeV
	Pair production	Bethe - Heitler ⁷	G4PairProductionRelModel	GammaInteractionConversion	$2m_0c^2$ - 100 TeV
	Rayleigh scat.	Livermore	G4LivermoreRayleighModel	not considered to be covered at the moment	100 keV - 100 TeV

^{*}Energy loss fluctuation corresponding to G4UniversalFluctuation model in Geant4-11.p01 also implemented for e+, e-

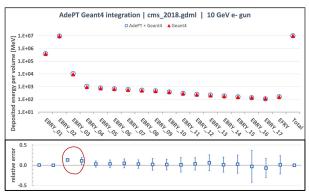
Correctness checks

- Validation against Geant4 standalone is essential
 - Comparisons to CPU references (in general Geant4-based) done for each added functionality
 - Both for standalone and Geant4 integration examples
- EM physics now fully validated
 - At ‰ level in the sampling calorimeter test case
- Still working on the last bugs/features in a hybrid workflow steered by Geant4

Sampling calorimeter example

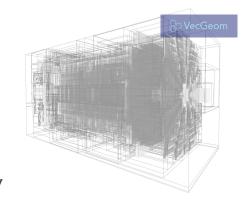


AdePT integration with Geant4



GPU geometry: VecGeom

- First implementation of GPU support few years old
 - C++ types re-compiled using *nvcc* in a separate namespace/library
 - In AdePT we wrote a custom global navigation layer calling lower level VecGeom APIs
- Improving gradually GPU support
 - Developed custom optimised navigation state, single-precision support
 - Moving from a simple "loop" navigator to an optimized BVH navigator
 - Adopting modern CMake GPU support
- Moving forward: specializing the VecGeom GPU navigation support
 - Portable less complex code, creating a surface-based view on device
 - An initial <u>prototype</u> is now being discussed



Parallelization in AdePT

- Simulation is done in steps, moving particles to either boundaries or physics processes
- All active tracks available are stepped at once (Geant4 transports one particle at a time)
 - Much higher degree of parallelism and more uniform work for the GPU
- No "thread-local" state, everything embedded in the track
 - Energy, position/direction, state needed across steps
 - Random number generator state (RANLUX++) per track to ensure reproducibility
 - Strategy to spawn a new sequence for daughter particles from the current state
- Tracks pre-allocated per particle type in thread-safe containers
 - Atomic counter to hand on track slots to be filled by kernels (explained later)

Track representation / access pattern



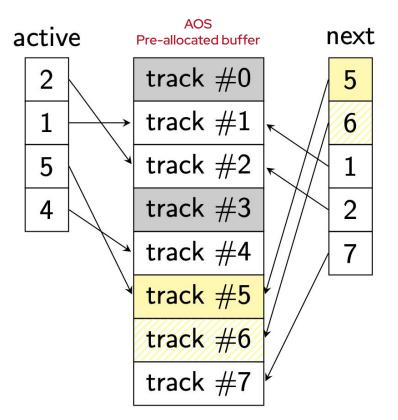
- <u>Investigated</u> entity component systems approach used in gaming (SoA)
 - Now also investigating track data structure transformations using



- No definitive conclusion yet
 - In realistic setups track data access is just a fraction of the loads/stores
 - Making small kernels accessing just part of the data introduces other overheads
- Difficult to implement coalesced memory access in simulation
 - Complex and sparse (accessed) data models. Geometry is a pathologic case.
 - The stochastic nature of the problem destroys locality.
 - Killed tracks leave random holes in the track data structure.
 - Placing statically data makes accessing it look like a "whack-a-mole" game
 - We need data regrouping solutions

Handling arrays of tracks in AdePT

- Store indices of active tracks (per particle type)
 - Parallelize transportation kernels over these indices
- Queue indices for "next" active tracks
 - Both secondaries and "surviving" tracks
 - Implemented with atomic counter
- Run transportation kernel stepping the active tracks
 - Here track #1, #2 and #5 survive, track #4 dies,
 and track #6 and #7 are produced
- Swap active
 → next before next iteration
 - Compacting unused slots now possible



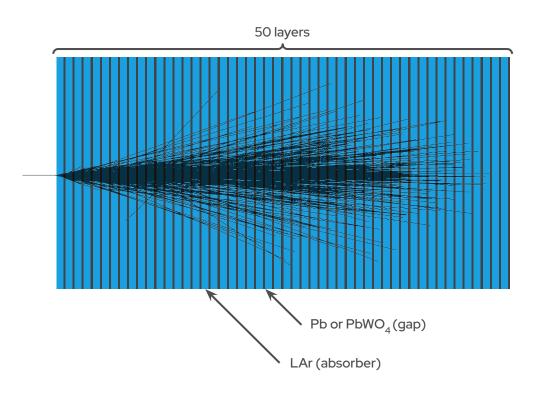
Stepping workflow, a first approach



- Can start kernels for particle types in parallel streams (transport is independent)
- Synchronization means overhead
 - Synchronize on the GPU via CUDA events
 - Synchronize with host once at the end of the step (stepping loop control, transfer hits)
- Main optimization playground
 - Better work balancing between warps, reducing impact of tails, better device occupancy

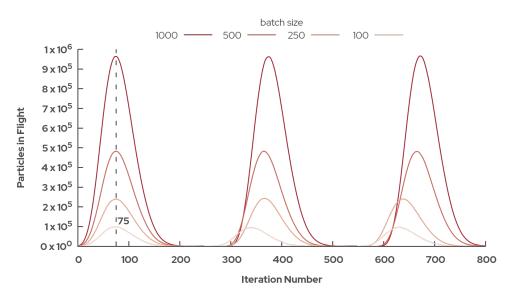
Simplified Calorimeter Benchmark

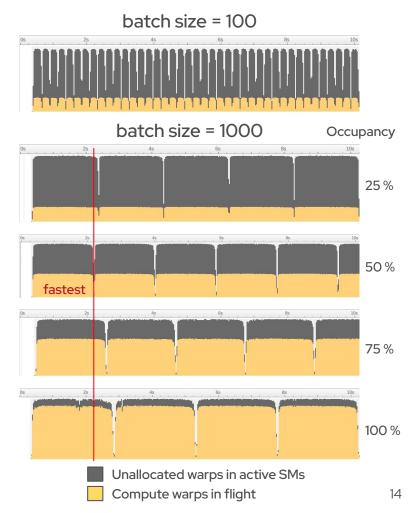
- Calorimeter with 50 layers
- Simulate 10,000 particles
 - 10 GeV electrons as primaries
- Configuration parameter space
 - Number of particles per batch
 - Number of registers per thread
 - Number of threads per block
- Compare on different hardware
 - Nvidia RTX 2070
 - Nvidia RTX 8000
 - Nvidia Tesla V100S



Run Time Characteristics

- putting more work per batch does more work in the same #iterations (steps)
 - o limited by available memory AND available tracks
- hints already to using strategies to fill the gaps
 - e.g. more CPU threads doing concurrent events





Kernel Launch Configurations

- ▶ 1024 Threads / SM
 - 4 schedulers x 8 warps/scheduler x 32 threads/warp
- ► 65536 Registers / SM
 - 4 register files x 16384 registers
 - 1 float = 1 register, 1 double = 2 registers
- 96 KB L1 Data Cache / Shared Memory
- Theoretical Occupancy (-maxrregcount or __launch_bounds__)
 - 256 regs/thread (256 threads, 8 warps) \Rightarrow 25%
 - 160 regs/thread (320 threads, 10 warps) \Rightarrow 38%
 - 128 regs/thread (512 threads, 16 warps) \Rightarrow 50%
 - 96 regs/thread (640 threads, 20 warps) \Rightarrow 63%
 - 80 regs/thread (768 threads, 24 warps) \Rightarrow 75%
 - 64 regs/thread (1024 threads, 32 warps) ⇒ 100%

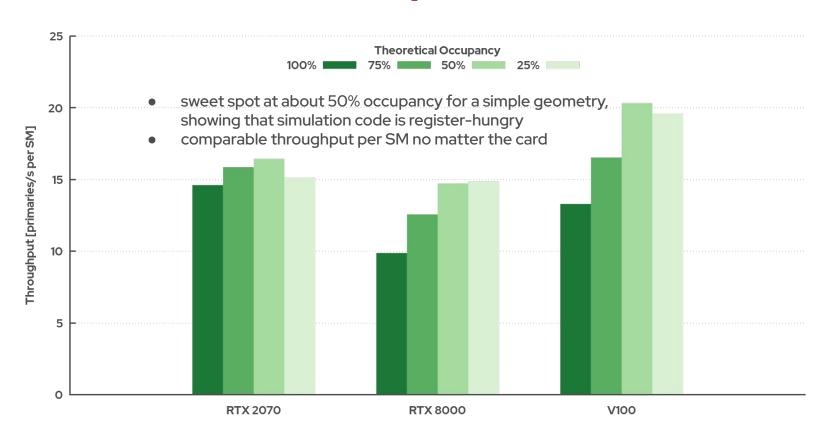
Higher parallelism

Factor Threads

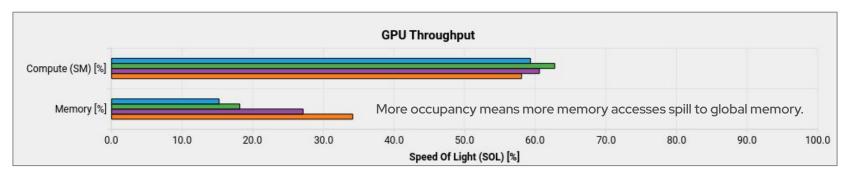
Turing SM

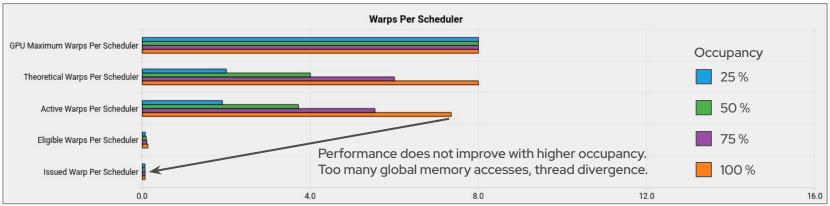


Relative Performance per SM

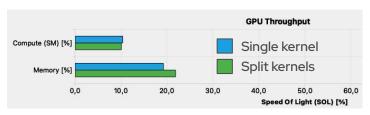


GPU Throughput (RTX 2070)





Case Study: Thread Divergence



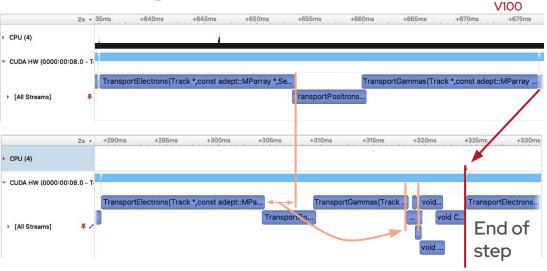
Problem: Threads in transport kernels diverge because of diverging interactions

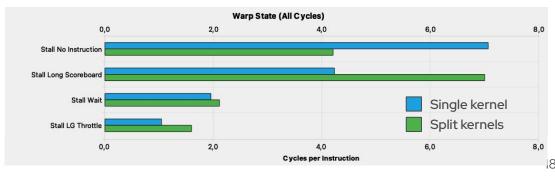
→ 13 / 32 threads active on average

Here: Split off interaction computations from cross-section and geometry kernels (one kernel for pair creation, one for ionisation, ...)

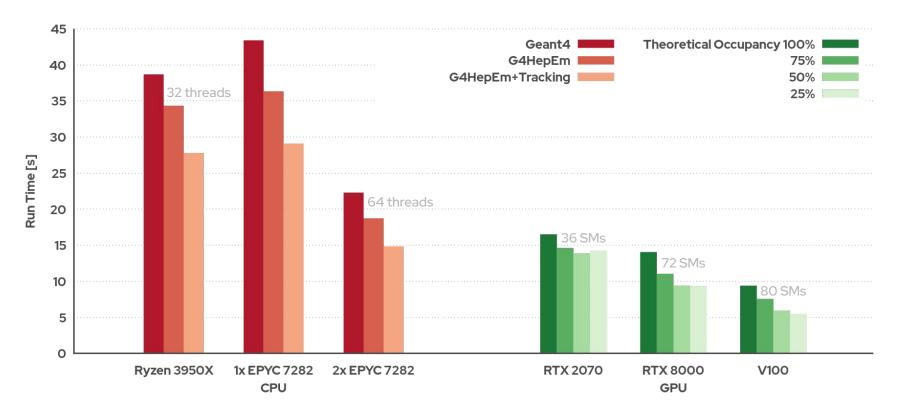
Result: 17 / 32 threads active for physics + geo 29 / 32 threads active for Bremsstr. Run time: $6.4 \text{ s} \rightarrow 5.5 \text{ s}$

Conclusion: Keeping threads coherent is key for detector simulation Generally difficult; stochastic processes





CPU vs GPU Performance



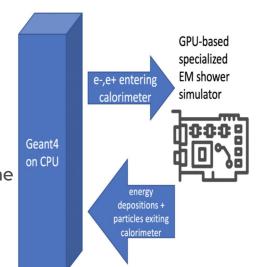
Performance portability

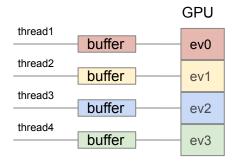


- oneAdePT port to oneAPI of an AdePT snapshot
 - core utilities, magnetic field, RNG, G4HepEM
 - No way around calling legacy CUDA code compiled in VecGeom
- Many obstacles for migrating CUDA to DPC++ code
 - SYCL limitations in calling virtual functions or function pointers, non-const globals, support for std:: math functions, support for CUDA compiled libraries, documentation
- Triggered investigations and work in VecGeom
 - Non-virtual dispatch and CUDA compilation using clang, deeper restructuring needed
 - Specializing geometry for GPU needed for both portability and better performance
 - Further efforts for portability postponed until solving this blocker

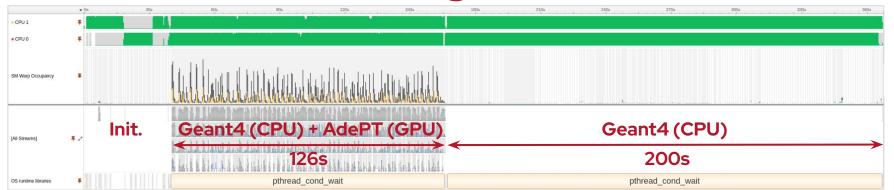
AdePT-Geant4 integration

- AdePT only provides EM physics for e⁺, e⁻ and γ
 - Cannot be used standalone for simulating a full experiment
 - In a first phase it could be used as accelerator for the EM part, in the same way as fast simulation models can be used in Geant4
- Developed an integration interface allowing a Geant4 region to become the "GPU region"
 - Intercepting and buffering for GPU particles sent asynchronously by Geant4 threads
 - May be in future applicable to the full detector, handing produced hadrons back to Geant4





CMS Simulations: Integrated and Standalone

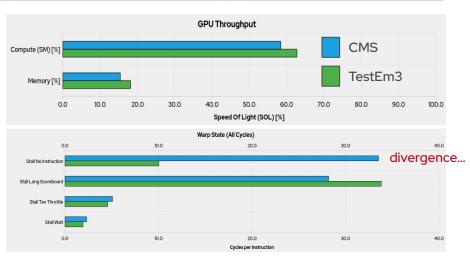


Integrate into Geant 4?

Above is a timeline of a simulation of CMS comparing the AdePT integration and Geant4 (Ryzen 3950X, RTX2070), with a **speedup of 37%** when using 2 CPU threads + 1 GPU vs only 2 CPU threads.

Impact of detector geometry?

On the right, 10^6 electrons at E=10 GeV on an Nvidia Tesla V100 with TestEm3 geometry vs the CMS geometry. The total simulation run time for the simplified calorimeter (TestEm3) setup is **549s** vs **1455 s** for the CMS geometry (a slowdown of 2.65x).



User actions & scoring

- Geant4 calls user code for performing custom run, event and stepping actions
 - Should we provide the same for simulation running on GPU?
 - Do we have to run (complex) user code there?
 - Code efficiency, device data management, transfer to host
- Solving this was not an immediate priority
 - First target: EM calorimeters, allowing for pre-defined scoring type
 - However simple energy deposits are not enough in several use cases
- A simplified approach based on static binding possible at this stage (we compile the transport kernels)
 - Init on device, score energy deposits, copy hits to host, clear
 - Called for sensitive detectors within the device stepping loop

```
SimpleScoring.h
struct SimpleS
 BasicScoring *InitializeOnGPU();
 device void Score(params);
 template <typename Stream>
 void CopyHitsToHost(Stream
                 &stream)
 void ClearGPU(Stream &stream)
};
using AdeptScoring = SimpleS
electrons.cuh
template <typename Scoring>
 global void
TransportElectrons(Scoring *s)
 s->Score(track state pars);
```

Outlook

- A challenging project, the problem is far from a perfect match for GPU
 - Fast progress due to some code refactoring done before AdePT (VecGeom, field)
 - Re-writing these is now necessary due to performance reasons
 - Several performance limitations in the path still to be addressed, some require deep code restructuring
- Prototypes for standalone and Geant4-integrated workflows available
 - Realistic examples for LHC setups, GPUs can be used in a Geant4 native application
 - Optimization work ongoing, performance not yet on a GPU-efficient baseline
- Most initial AdePT objectives complete
 - Still to decide on the strategy for larger developments and more efficient integration with the experiment's simulation code
- Collaborating on common development topics with the Orange team is essential: geometry, integration with Geant4 and experiments code, ...