

# High Energy Physics Center for Computational Excellence Detector Simulation

**Tom Evans  
for Celeritas, HEP-CCE**

**DOE HEP CCE All Hands  
Dec 18 2023**

# HEP-CCE Phase 2 – Detector Simulation

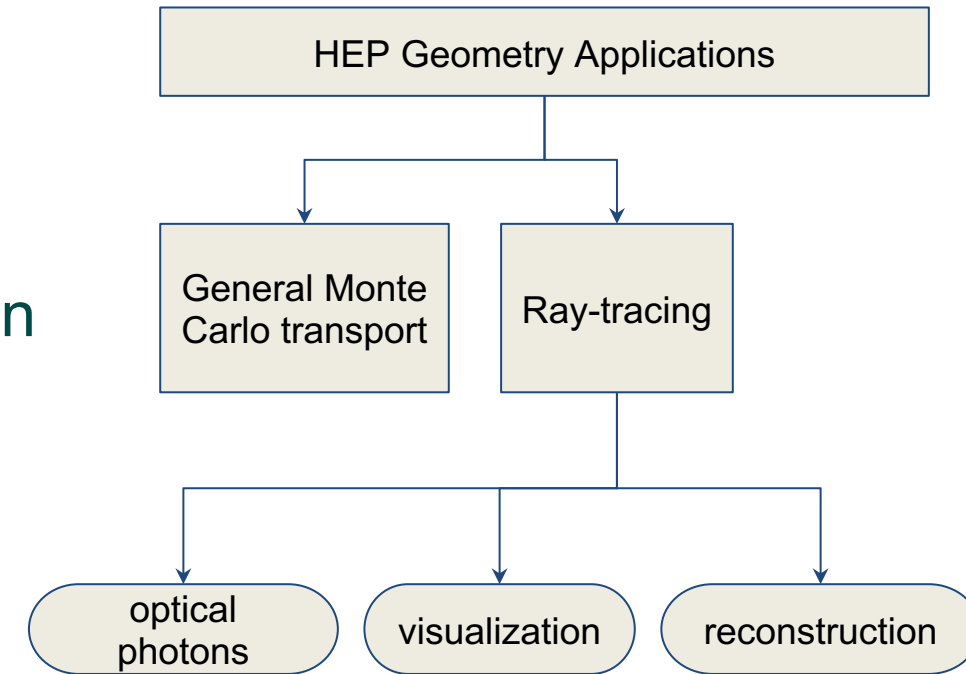
- Platform portable geometric tracking
- Optical photons

Specific objectives and deliverables discussed this afternoon

# Geometric capabilities - current status

- Monte Carlo general particle simulations use **VecGeom** as the standard package
  - Adapted from CPU vectorization prototype
  - Extended to work for Nvidia CUDA
  - Supports Geant4 model construction and tracking
- **Optical photons use several approaches**
  - Native Geant4 navigation in full event-stepping loop
  - Opticks HEP library (uses Nvidia OptiX™ ray-tracing engine)
- **Drawbacks:**
  - No platform portability - existing libraries are Nvidia-centric
  - Distinct code bases reduce opportunities to amortize common capabilities
- **Objective:**
  - Support platform portability across DOE LCFs
  - Unified geometry components that can support multiple applications

- **General Monte Carlo navigation**
  - Robust boundary crossings (double precision)
  - Error checking
  - Graph-based search and subcell optimizations
- **Ray-tracing for optical photons and visualization**
  - Graph-based multi-surface tracking
  - Track batching
  - Variable-precision
- **Commonalities:**
  - Portability: (Intel, AMD, Nvidia GPUs, multicore)
  - Physics-based simulations require high-precision tracking
  - Bounding volume hierarchy (BVH) graph-based optimizations
  - Many low-level components can be shared
    - High-level APIs will be different depending on application

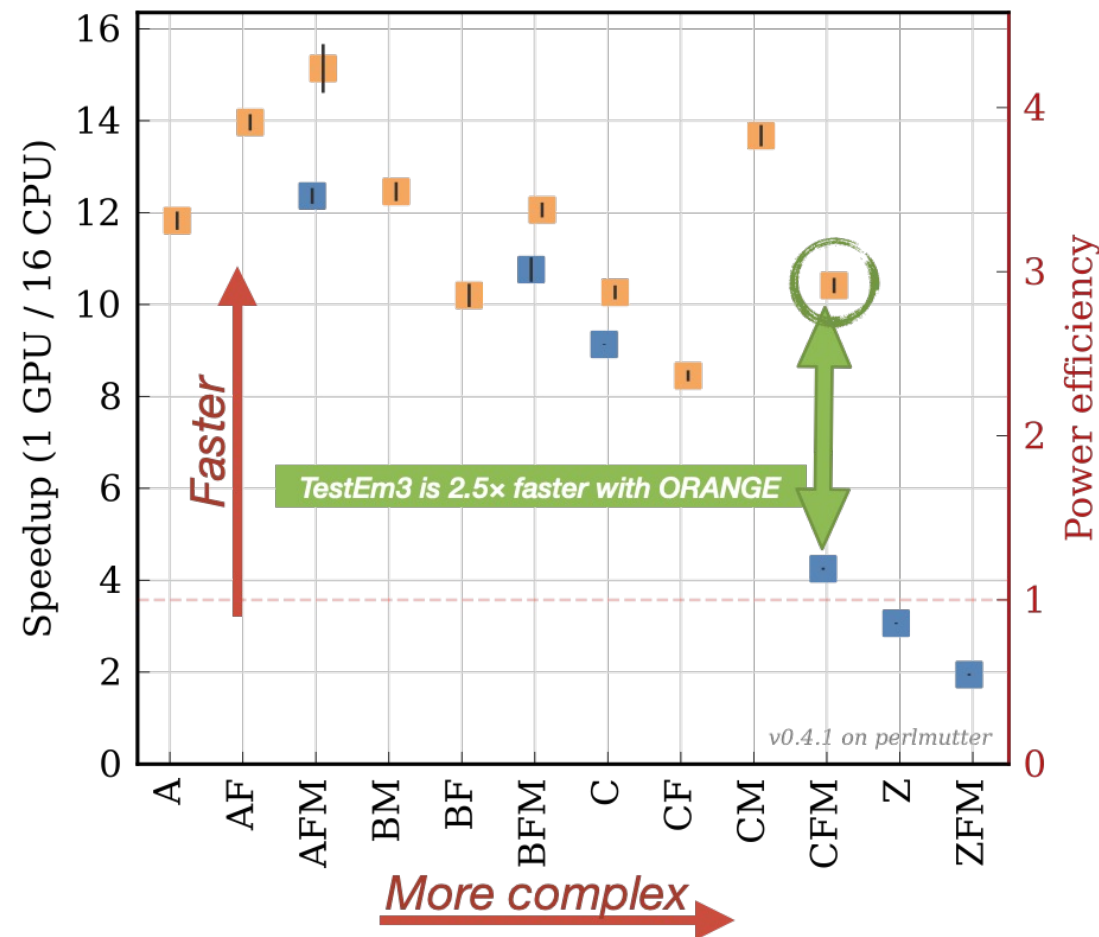


**Geometry packages must support existing HEP model formats and descriptions**

## Geometry - benefits to HEP community

- Provide geometric capabilities for spectrum of applications
  - Monte Carlo detector simulation
  - Optical photon ray-tracing
  - Reconstruction tracking
  - Visualization
- Platform portability
  - Support current and future architectures in DOE LCFs
- Community sustainability
  - Interdisciplinary applications beyond HEP
- **ORANGE** geometry engine
  - New engine developed in Celeritas
  - Tracking based on CSG tree of surfaces **constructed from volumes**
  - Collaboration to implement surface-based tracking in **VecGeom**

Problem definition		Modifier	
A	"infinite" medium	F	+field
B	simple-cms	M	+msc
C	idealized calorimeter		
Z	cms2018		



# Optical photon transport - current status

- Critical application spaces:
  - DUNE (art)
  - LZ (BACCARAT)
  - CalVision
- Current approaches:
  - Direct transport using Geant4 generalized navigation engine
    - e.g., BACCARAT spends >95% of CPU time on optical photon transport when using Geant4
  - Full transport replaced with *fast* approximation methods
  - Optimized optical photon transport using Opticks
- Drawbacks:
  - Portability
    - Geant4 capabilities CPU-only
    - Opticks only available on Nvidia architectures
  - Performance for direct simulation

# Optical photon transport - application objectives

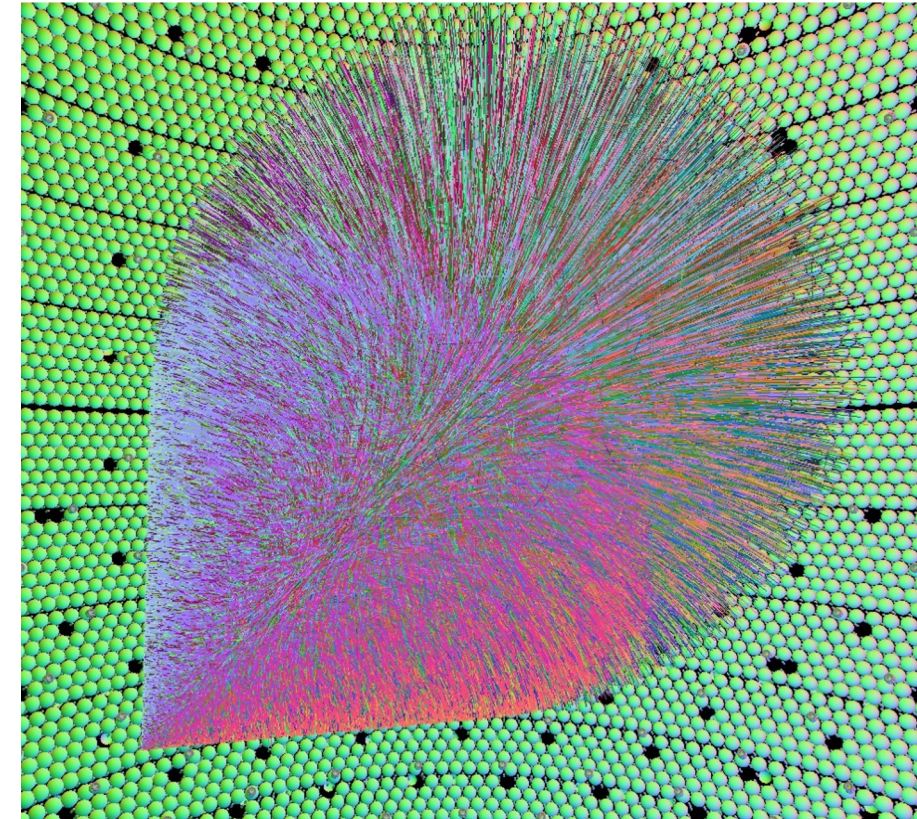
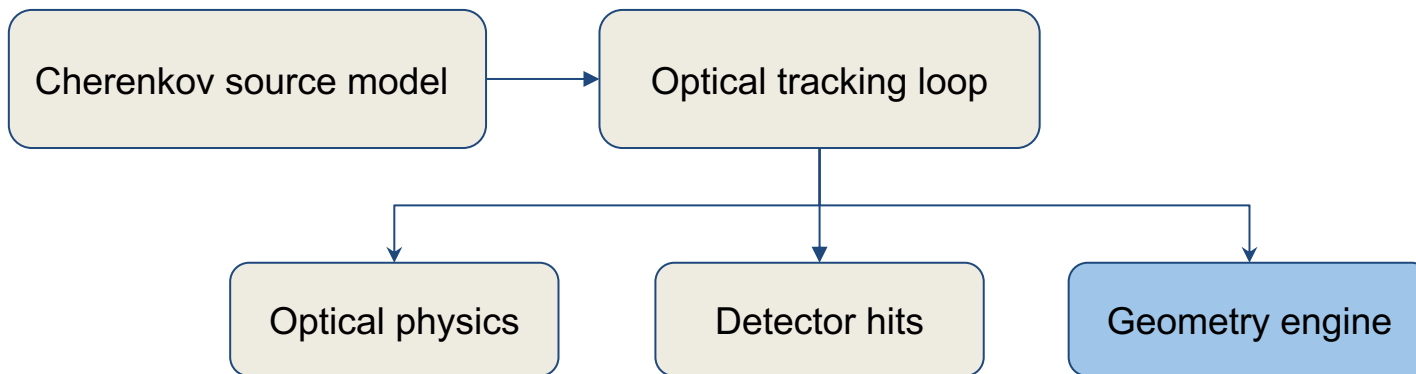
- Fast throughput for direct transport
  - Avoid or accelerate fast approximations (e.g. optical maps in BACCARAT)
- Platform portability
- Ease of integration into existing application frameworks
  - art
  - BACCARAT
  - future luminescent detector applications
- Compatibility with Geant4 geometry models
  - Large detector regions
  - Large detector arrays

Optical photon transport efficiency is closely tied to geometric modeling capabilities



# Optical photon transport approaches

- Opticks demonstrated that very high efficiencies can be obtained using GPU ray-tracing
  - >1000x Geant4 direct simulation
- Efficiency gains result from:
  - Specialized optical photon ray-tracing loop
  - Hardware acceleration (BVH traversal)



Optical photon shower (JUNO neutrino detector); S.C. Blyth, *Opticks: GPU Optical Photon Simulation for Particle Physics with Nvidia OptiX™*, CHEP 2018.

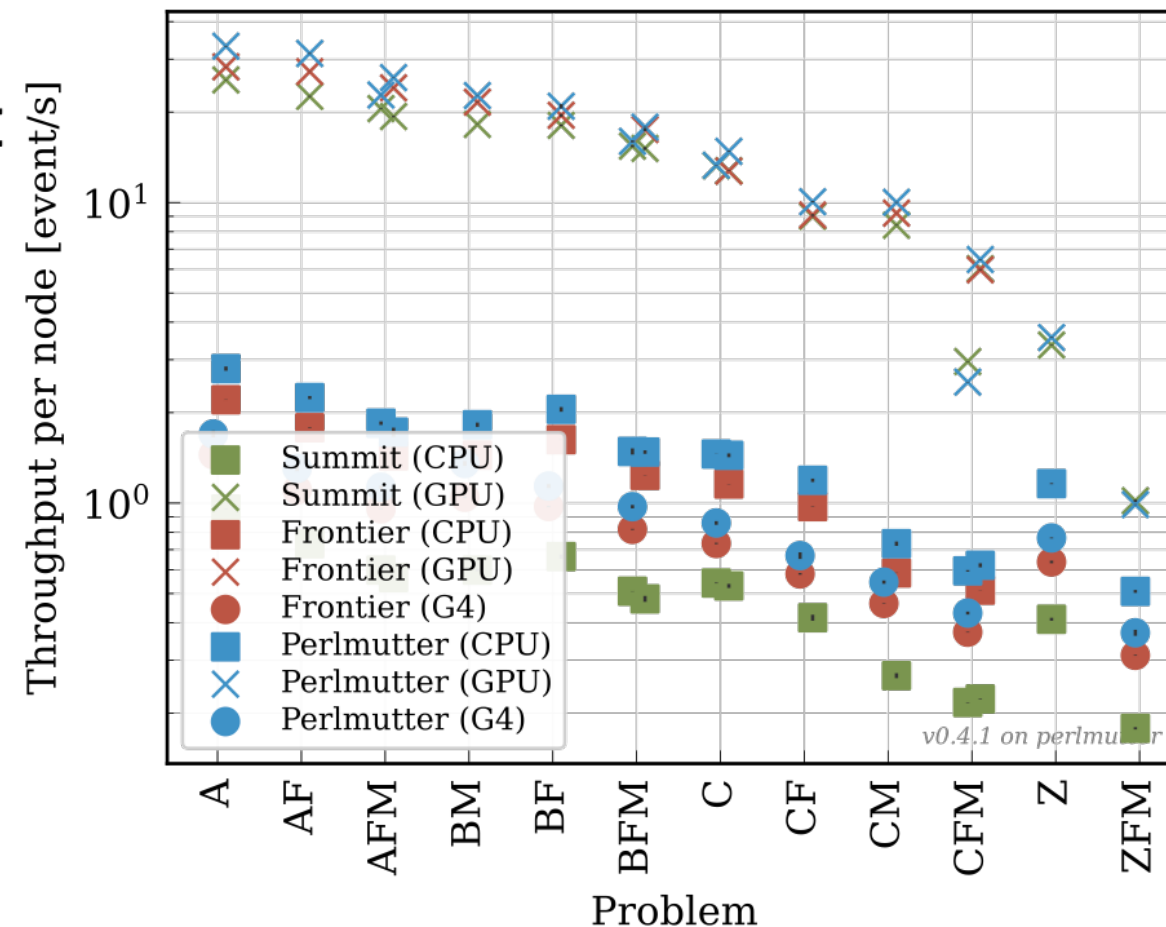


# Detector simulation throughput

- GPUs *cannot* be ignored if present
- AI/ML “revolution” guarantees more coprocessors at all scales

Per-node stats for DOE supercomputers

Machine	Arch	Card	TDP (W)	Cores*	Cards
Summit	CPU	IBM Power9	190	‡22	2
	GPU	Nvidia V100	250	80	6
Perlmutter	CPU	AMD EPYC 7763	280	64	1
	GPU	Nvidia A100	250	108	4
Frontier	CPU	AMD EPYC 7453	225	‡64	1
	GPU	AMD MI250x	500	‡220	‡4



\*or SMs;

‡Each card has 2 GPUs

‡One core reserved per GPU

EM only, no SDs