



U.S. DEPARTMENT OF
ENERGY

Office of
Science

Detector Simulation on Modern Processors

Vectorization of Physics Models

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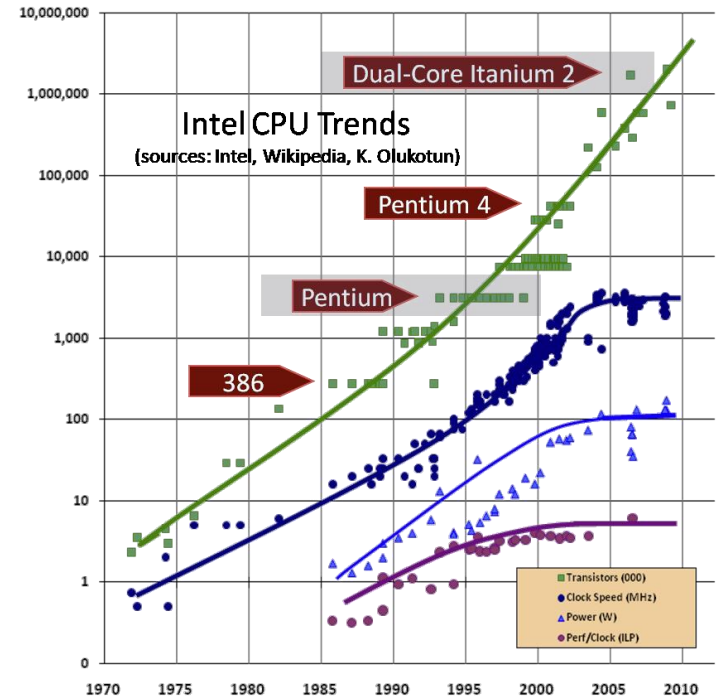
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Contents

- Introduction
- GeantV
- Vector Physics Model
- Validation and Performance
- Conclusion

Introduction

- Motivations
 - Performance of our code scales with clock cycle
 - HEP code needs to exploit new architectures to improve
 - Data & instruction locality and vectorisation
 - Portability, better physics and optimization will be the targets



Introduction

- GeantV Goals

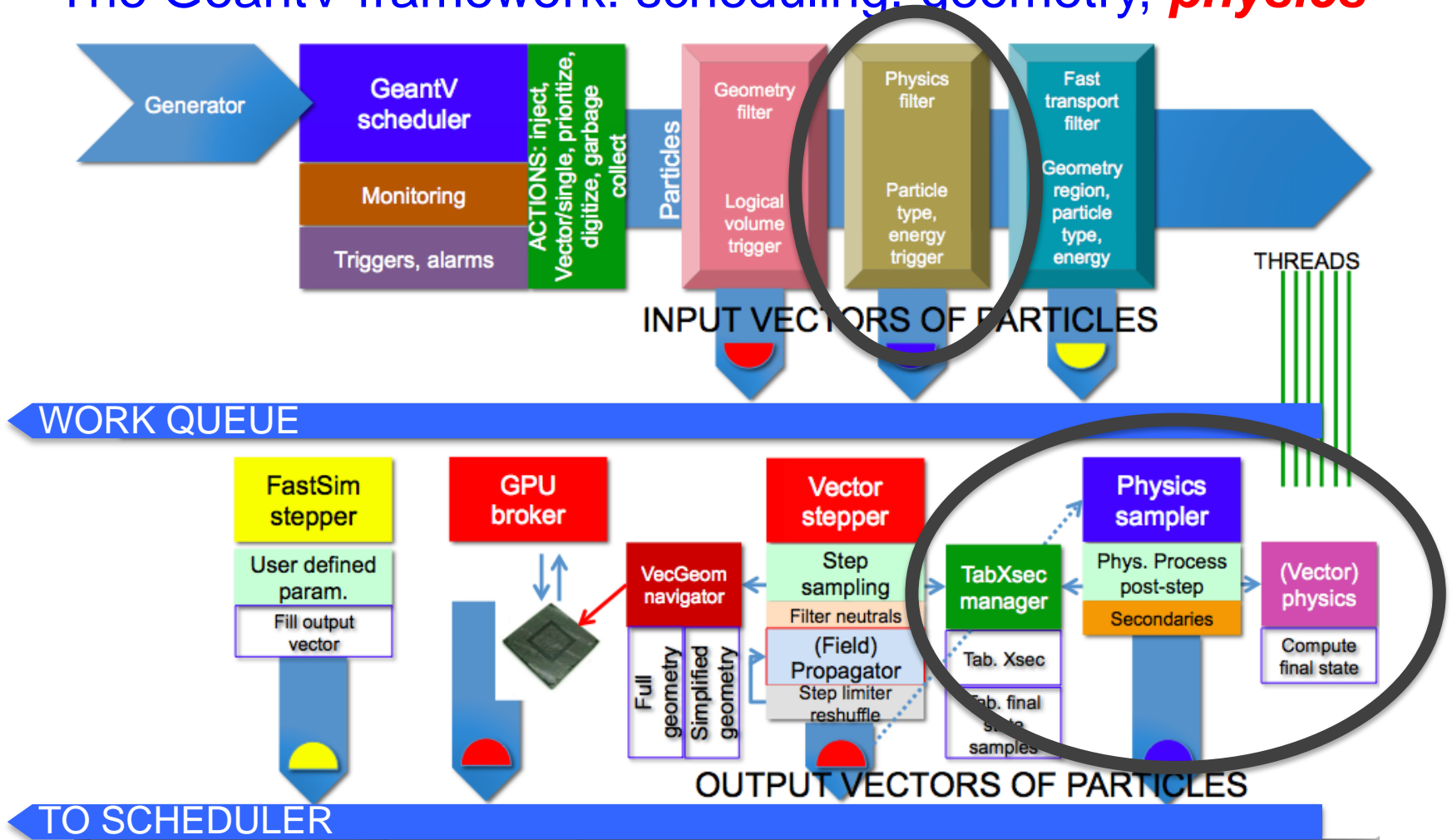
- Develop an all-particle transport simulation program
 - 2 to 5 times faster than Geant4
 - Continues improvement of physics
 - Full simulation and various options for fast simulation
 - Portable on different architectures, including accelerators (GPUs and Xeon Phi's)
- Understand the limiting factors for 10x improvement



See [The GeantV project: preparing the future of simulation](#)
on **14 Apr 2015** at **17:15**

GeantV: The next generation detector simulation toolkits

- The GeantV framework: scheduling, geometry, *physics*



Vector Physics Model

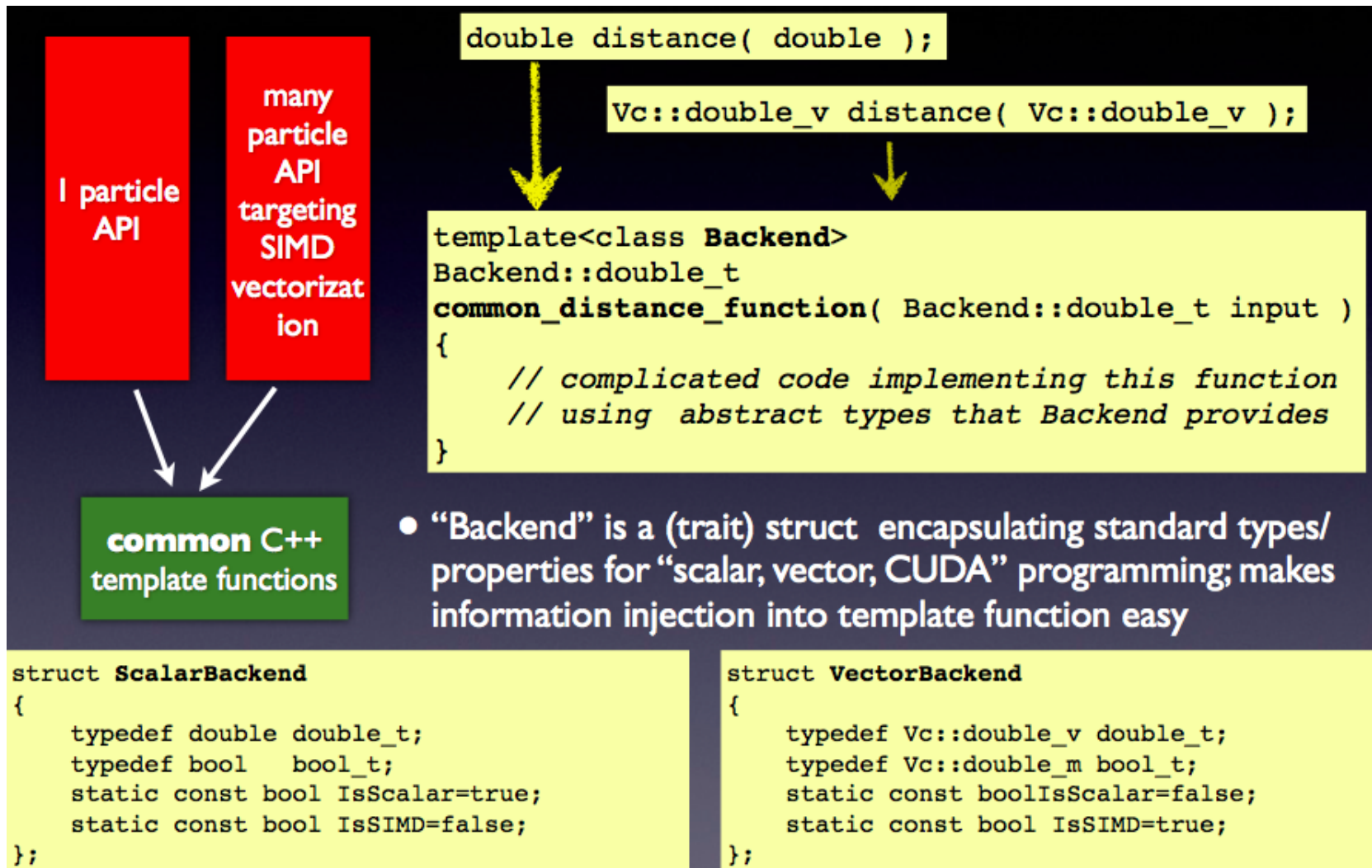
- Assumption: particles are independent during tracking
- Vectorization of the density of collisions, Ψ

$$\Psi = \sum_{k=1}^{\text{ntracks}} \Psi_k(\text{sequential iteration}) \rightarrow \prod_{k=1}^{\text{ntasks}} \Psi_k(\text{task-based execution})$$

$$\Psi_k(\vec{r}, \vec{v}) = \int d\vec{r}' [S_k(\vec{r}', \vec{v}) + \int \Psi_k(\vec{r}', \vec{v}') I_k(\vec{r}', \vec{v}' \rightarrow \vec{v}) d\vec{r}'] T_k(\vec{r}' \rightarrow \vec{r}, \vec{v})$$

- S_k (source): vector scheduler
 - T_k (transport): VectGeom + vector propagator (geometry limited step)
 - I_k (interaction): **vectorized physics** (step length, secondary production, ...)
- **Vector strategies: data locality and instruction throughput**
 - decomposition sequential tracking and regroup them by tasks
 - algorithmic vectorization and parallel data patterns
 - targeting both external and internal (SIMD) vectorization

Portability (Template Approach): Scalar, Vector, CUDA, MIC

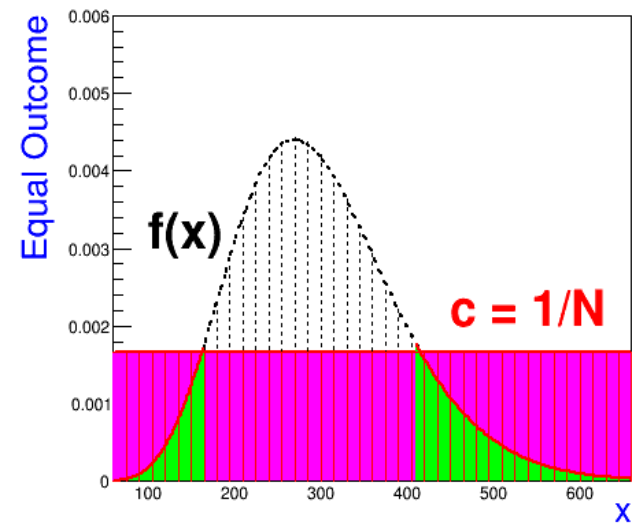
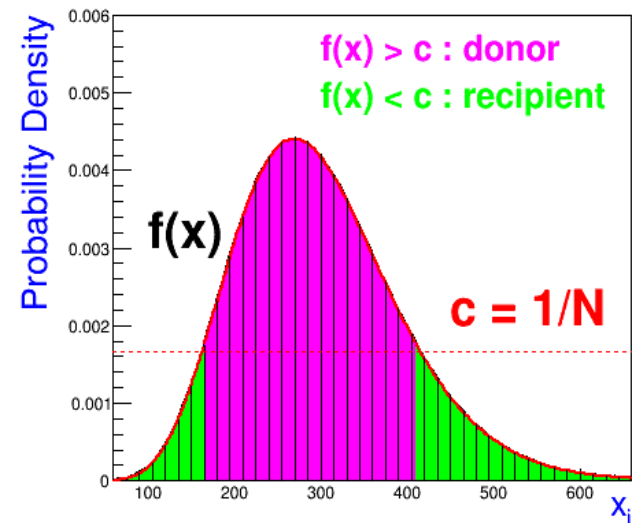


Prerequisites to Achieve Efficient Vectorization

- Vectorized pseudo-random number generator
- Data layout: coalesced memory access on vector operands
 - SoA (struct of array) tracks parameters ($\mathbf{x}, \mathbf{p}, t, E \dots$)
 - ordered and aligned data arrays
- Data locality for the vector of particles
 - particle type, geometry and material, physics process
- Vector operations
 - identical instructions on each components of the vector
 - no conditional branches, no data dependencies
 - replace non-vectorizable algorithms (ex. composition and rejection methods) by alternatives

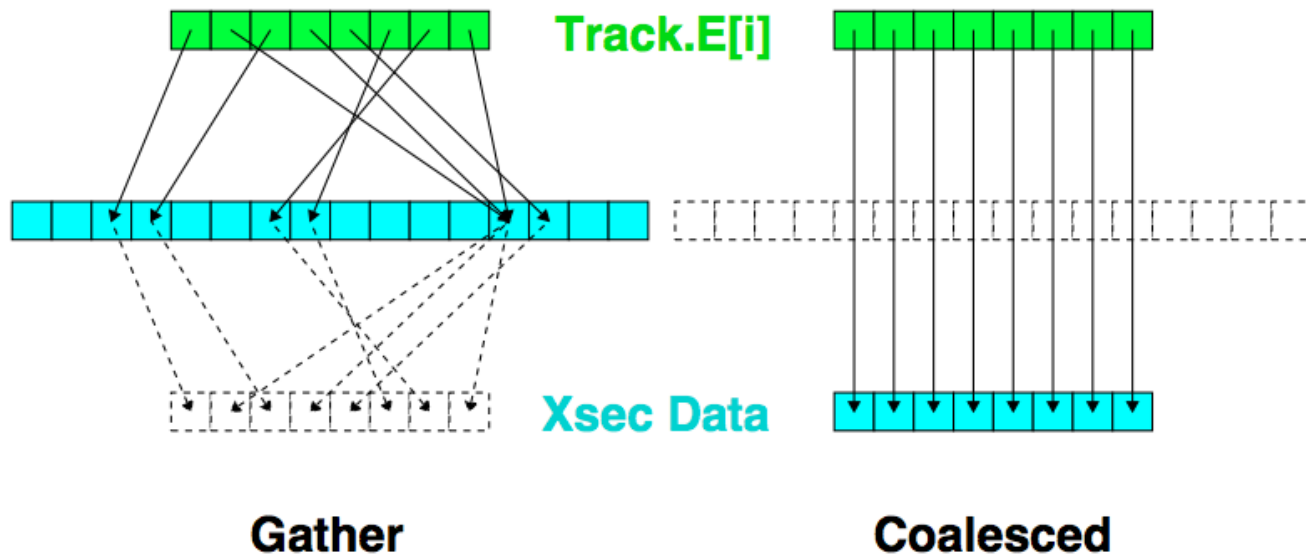
Sampling Secondary Particles: Alias Method (A.J.Walker)

- Recast a cross section, $f(x)$ to N equal probable events, each with likelihood $c = 1/N$
- Alias table
 - $a[\text{recipient}] = \text{donor}$
 - $q[N] = \text{non-alias probability}$
- Sampling x_j : random u_1, u_2
 - bin index: $N \times u_1 = i + \alpha$
 - sample $j = (q[i] < u_2) ? i : a[i]$
 - $x_j = [\alpha j + (1-\alpha)(j+1)]\delta x$
- Replace composition and rejection methods (conditional branches – not vectorizable)



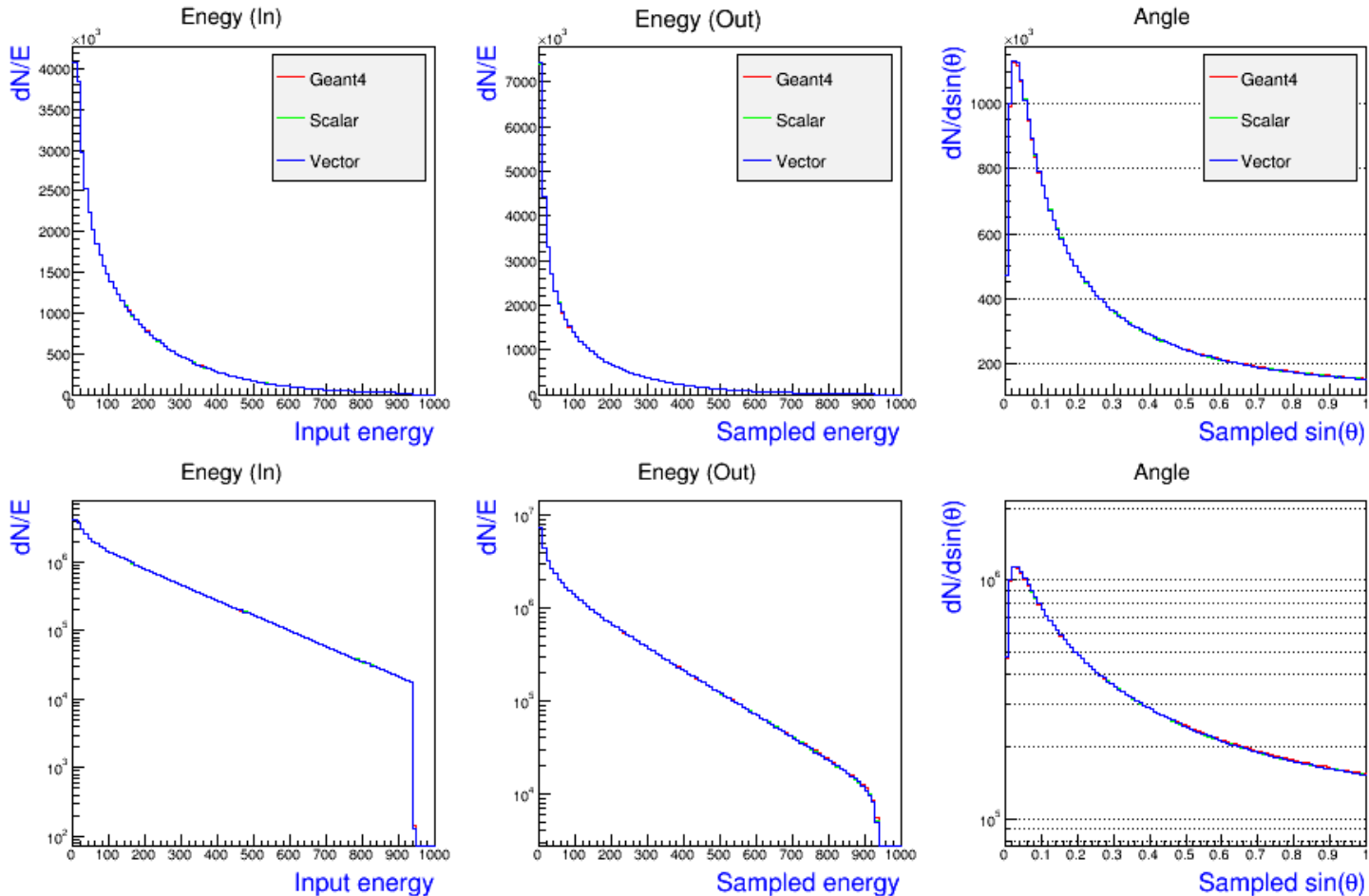
Coalesced Memory Access

- Sampling the step length and the physics process
 - cross section calculation on-the-fly (fully vectorizable, likely expensive)
 - tabulated physics (table-lookups, bandwidth limited)
- Gather data to enable contiguously ordered accesses
 - loss by overhead < gain by vectorization

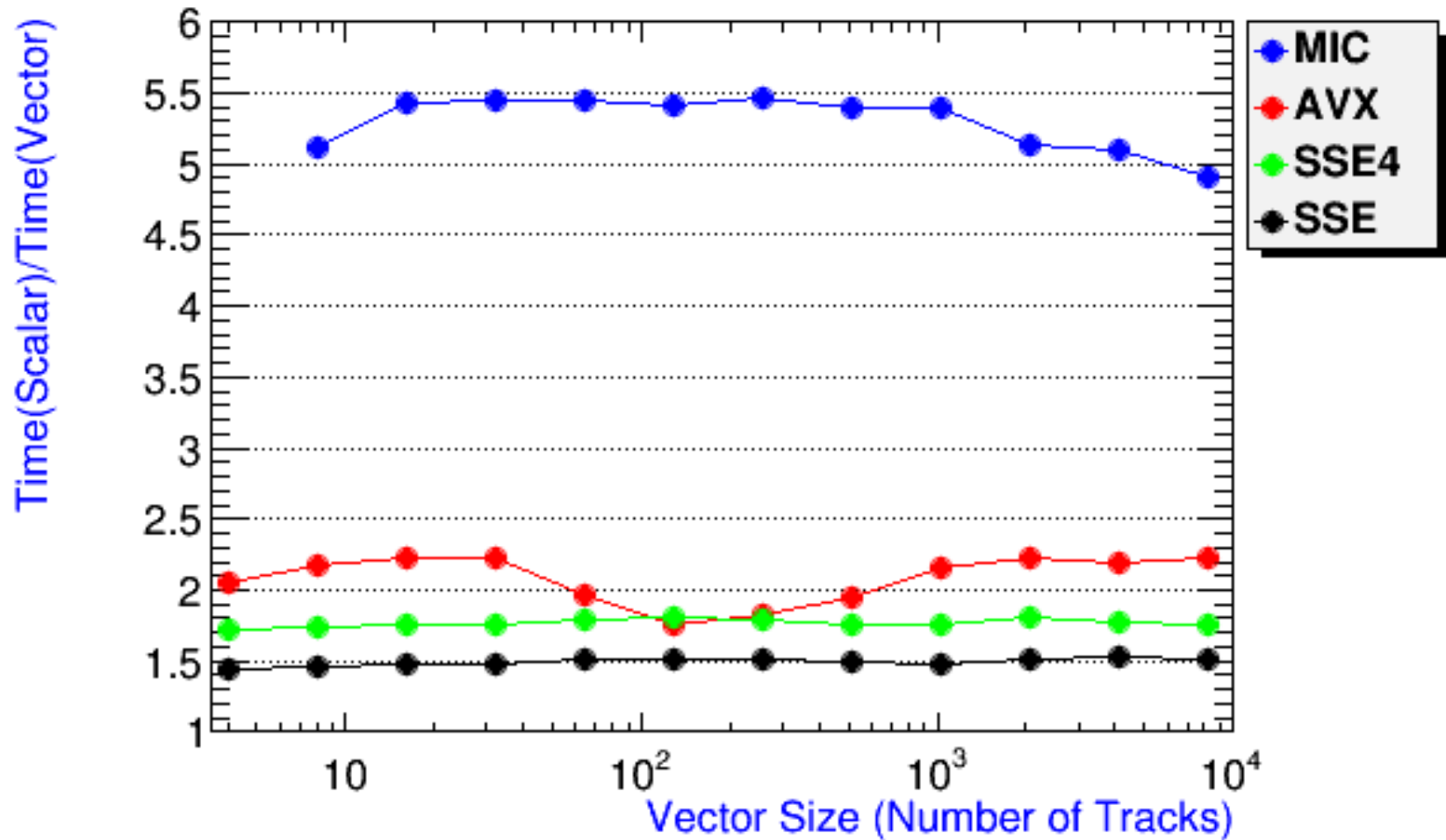


Validation: Alias vs. Composition and Rejection Method

- Compton (Klein-Nishina model): energy and angle of scattered photons



Vector Speedup: Factor 2 on Xeon



Runtime Performance

- Relative performance for sampling the secondary electron
 - Composition Method, Scalar, Vector
 - average time for 100 trials for 4992x100 tracks – SSE
 - Table size [input energy bins, sample energy bins]

Table size	Time with [100,100]	Time with [100,1000]
Composition Method	11.609	11.347
Alias Method, Scalar	8.439	10.080
Alias Method, Vector	5.446	6.185

- Note that Composition Method Klein-Nishina model is one of the most efficient composition and rejection examples ($\epsilon \sim 1$)

Status and Plan

- Implement one fully vectorized EM physics model (Klein Nishina Compton) and test with GeantV
 - Backend: Scalar, Vector, CUDA
 - Performance evaluation and validation
- Complete all EM physics by Dec. 2015

Primary	Process	Model	Secondaries	Survivor
e^-	Bremsstrahlung	SeltzerBerger	γ	e^-
	Ionization	MollerBhabhaModel	e^-	e^-
	Multiple Scattering	UrbanMscModel95	-	e^-
γ	Compton Scattering	KleinNishinaCompton	e^-	γ
	Photo Electric Effect	PEEffectFluoModel	e^-	-
	Gamma Conversion	BetheHeitlerModel	$e^- e^+$	-

- Extend for hadron physics and explore other algorithms

Conclusion

- Significant performance improvement achievable in detector simulation physics code using a combination of:
 - Alternative algorithm (reducing branching, etc.)
 - Vectorization
 - Increased use of code and data caches
- Using template techniques, code is portable to different modern computing architectures while still being tuned for each architecture.

Backup Slides

