

### Portable HPC for High-Performance Simulation



Federico Carminati Future of Many-Core Computing in Science 29 May 2014 Wigner Datacenter

## Thanks to



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### A luminous future for HEP...

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2009	Start of LHC - 2009: √s = 900 GeV	
2010 2011 2012	Run 1: $\sqrt{s} = 7-8$ TeV, L = 2-7 x 10 <sup>33</sup> cm <sup>-2</sup> s <sup>-1</sup> Bunch spacing: 75/50/25 ns (25 ns tests 2011; 2012 )	-25 tb '
2013	LHC shutdown to prepare for design energy and nominal luminosity	LS1
2015 2016	Run 2: $\sqrt{s} = 13-14$ TeV, L = 1 x $10^{34}$ cm <sup>-2</sup> s <sup>-1</sup> Bunch spacing: 25 ns	>50 fb
2017	Injector and LHC Phase-I upgrade to go to ultimate luminosity	S2
2019 2020 2021	Run 3: $\sqrt{s} = 14$ TeV, L = 2 x 10 <sup>34</sup> cm <sup>-2</sup> s <sup>-1</sup> Bunch spacing: 25 ns	-300 fb <sup>-1</sup>
2022	High-luminosity LHC (HL-LHC), crab cavities, lumi levelling,	
 2030	Run 4: $\sqrt{s} = 14$ TeV, L = 5 x 10 <sup>34</sup> cm <sup>-2</sup> s <sup>-1</sup> Bunch spacing: 25 ns	-3000 fb-1
		J. J.L.at

CERN openlab BoS – May 8, 2014



### A luminous future for HEP...



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# The Eight dimensions



- The "dimensions of performance"
  - Vectors
  - Instruction Pipelining
  - Instruction Level Parallelism (ILP)
  - Hardware threading
  - Clock frequency
  - Multi-core
  - Multi-socket
  - Multi-node

Micro-parallelism: gain
→ in throughput and in time-to-solution

Very little gain to be expected and no action to be taken

> Gain in memory footprint and time-to-solution but not in throughput

Possibly running different jobs as we do now is the best solution

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Expected limits on performance scaling						
	SIMD	ILP	HW			
THEORY	8	4	1.3			
OPTIMISED	6	1.57	1.2			
HEP	1	0.8	1.2			
Expected limits on performance scaling (multiplied)						
	SIMD	ILP	HW			
THEORY	8	32	43.			
OPTIMISED	6	9.43	11.79			
HEP	1	0.8				

- → in throughput and in time-to-solution
- Very little gain to be expected and no action to be taken
  - Gain in memory footprint and time-to-solution but not in throughput
  - Possibly running different jobs as we do now is the best solution

OpenLab@CHEP12



### Classical HEP transport is mostly local

 Event- or event tracklevel parallelism will better use resources but won't improve these points



(local)

- Material X-section tables
- Particle type physics processes



- Navigating very large data structures
- No locality
- OO abused: very deep instruction stack
- Cache misses



Deal with particles in parallel























A dispatcher thread puts particles back into transport buffers

> Everything happens asynchronously and in parallel

The challenge is to minimise locks

Keep long vectors

Particles are transported per thread and put in output buffers





A dispatcher thread puts particles back into transport buffers

> Everything happens asynchronously and in parallel

The challenge is to minimise locks

Keep long vectors

Avoid memory explosion

Particles are transported per thread and put in output buffers





### **Basket managers**

- One basket manager per volume
  - Receiving tracks entering the volume from generator or scheduler
  - Accessed by scheduler only
- Pool of empty baskets, one current basket + one basket for prioritized tracks
- Lock-free access for unique scheduler (only one thread can add tracks)
- Transportability threshold per manager
  - If threshold reached when adding tracks, the current basket is pushed in the work queue and replaced from the pool. Tracks added with the priority flag go to the priority basket which gets pushed to the priority side of the queue
  - Threshold(vol) = Ntracks\_in\_flight(vol)/2N\_threads rounded to %4 (min 4, max 256)



### Basket lifecycle



## Scheduling policies

- Workload balancing
  - Divide the work evenly to scale with number of workers
  - Queue control: garbage collection on work queue depletion
  - Improvement: schedule physics as separate task (process selection and discrete processes post-step)
- Memory management
  - Not active currently, the idea it to trigger hit/digits collection and memory cleanup on thresholds
- Keep large vectors
  - Raise transportability thresholds per volume
  - Postpone sparse tracks when not in garbage collection mode
- Trigger single track mode when vectorization gives just overhead

## Physics



- A lightweight physics for realistic shower development
- Select the major mechanisms
  - Bremsstrahlung, e+ annihilation, Compton, Decay, Delta ray, Elastic hadron, Inelastic hadron, Pair production, Photoelectric, Capture + dE/dx & MS
- Tabulate all x-secs (100 bins -> 90MB)
- Generate (10-50) final states (300kB per final state & element)
- Not good as Geant4, but it could be the seed of a fast simulation option
- Independent from the MonteCarlo that actually generates the tables





#### Recap of performance status

provided new optimized vector interfaces for some elementary solids and geometric base classes ( implemented important functions for particle navigation )

Soverall performance gain in a standard navigation benchmark ( in toy detector with 4 boxes, 3 tubes, 2 cones ) - comparison to ROOT/ 5.34.17



	16 particles	1024 particles
Intel IvyBridge (AVX)	~2.8x	~4.0x
Intel Haswell (AVX2)	~3.0x	~5.0x
Intel Xeon-Phi (AVX512)	~4.Ix	~4.8x

Xeon-Phi and Haswell benchmarks by CERN Openlab (Georgios Bitzes)

CHEP13 paper: http://arxiv.org/pdf/1312.0816.pdf

SIMD

distFromInside

mothervolume

Annual Concurrency Forum Meeting, 02-04-14

## Portable HPC?



- Straight "vectorisation" of existing code is difficult to impossible
- Resulting code is hard to read and maintain
  - And it is largely compiler-dependency
- Porting to different high end devices is very difficult
- Explored solution is to use template specification for solid placement, specialisation and code generation
- Highly optimised modular "codelets" à la STL are used to construct algorithms

## Solid specialisation



CreateTube(rmin, rmax, al1, al2)





Single particle interface

Vector interface External CUDA kernels

C-like abstract kernels





C-like abstract kernels





C-like abstract kernels





## Performance examples



- Performance of the inside method for a tube on an Intel CPU
- As you can see the performance is improved even for the scalar version

#### Unspecialized scalar

- Specialized scalar (rot, trans)
- Specialized vector (rot, trans)
- Specialized scalar (rot, trans, type)
- Specialized vector (rot, trans, type)
- ROOT

Usolids



#### Particle Transport on the GPU



#### **GPU Prototype: Three Core Components**





April 2, 2014

### **Physics Validation of GPU Physics**

Compare simulated physics outputs
RED BLUE



🗖 🛟 Fermilab

#### **Performance Evaluation**

• Hardware (host + device)

	Host (CPU)	Device (GPU)
<i>M2090</i>	AMD Opteron <sup>TM</sup> 6134 32 cores @ 2.4 GHz	Nvidia M2090 (Fermi) 512 cores @ 1.3 GHz,
<i>K20</i>	Intel® Xeon® E5-2620 24 cores @ 2.0 GHz	NVidia K20 (Kepler) 2496 cores @ 0.7GHz

Performance measurement



#### **Performance: Realistic Simulation**

- A simple calorimeter (CMS Ecal) with the CMS b-field map
- Tracking for one step: split kernels (GPIL+sorting+Dolt)

	CPU [ms]	GPU [ms]	CPU/GPU
AMD+M2090	748	37.8 (62.6)*	19.8 (11.9)*
Intel®+K20M	571	30.4 (81.9)*	18.7 (7.0)*

• Performance by each kernel (% of the total application time)



#### **Performance Issues and Considerations**

• Observed issues (by the Nvidia profiler)

Considerations



### **Challenges and New Strategies**

• HEP detector simulation (ex. Geant4) is a giant

Coprocessor architectures

• Top-down approach



### **Incorporating into the Vector prototype**

• The vector prototype started at CERN (talk by F. Carminati)

Integration to the vector prototype (GPU broker)



# CUDA in GeantV geometry



- Enable dispatching to CUDA as a co-processor
- Run separate to scalar and vectorized code
- Use same codebase
- Achieved by abstracted, templated algorithms



### Separate compilation of backends



## Shape measurements





- Scalar, vector and CUDA code templated from <u>same</u> abstracted algorithm
- Dispatch on the fly to optimal processor
- Typical GPU scaling; high minimum input threshold

## Where are we now?

- Scheduler
  - The new version, hopefully improved of the scheduler has been committed and we are testing it
- Solids
  - The proof or principle that we can achieve large speedups (3-5+) is there, however a lot of work lays ahead
- Navigator
  - "Percolating" vectors through the navigator is a difficult business. We have a simplified navigator that achieves that, but more work is needed here
- Physics
  - Can generate x-secs and final states and sample them, but there are still many points to be clarified with Geant4 experts





29 \*UGeom == USolid + navigation



29 \*UGeom == USolid + navigation





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FRI







29 \*UGeom == USolid + navigation







29 \*UGeom == USolid + navigation

## Conclusion



- Work on a full prototype is progressing
- The first performance figures are very encouraging
- The template specialisation technique seems to provide a reasonable model for portable HPC
- We hope to have a demonstrator by the end of the year