## Traccc: GPU Track Reconstruction Library

Beomki Yeo UC Berkeley and LBNL

On behalf of the traccc team





### Motivation for Hardware Accelerator in HEP Computing

- Hard to sustain computing budget model with traditional CPUs in future experiments (e.g. High Luminosity LHC)
- Hardware accelerators such as GPU are expected to outperform the CPUs



Image credit

#### What is GPU?

- Graphical Processing Unit
  - a.k.a graphics card
  - Major Vendors: NVIDIA, AMD, and Intel
- Invented for the visualization on PC screen
   Now also used as a hardware accelerator
- Two types of GPU
  - Integrated in CPU
  - Discrete GPU



Image credit



Discrete GPU Market Share

### Why GPU? (Performance per USD)



*Theoretically* GPU is faster and more economical than CPU

### Why GPU? (Performance per Watt)

Single (FP32)

**Double (FP64)** 



Theoretically GPU is faster and more economical than CPU

Beomki Yeo

### CPU vs GPU

#### • CPU

- Small number of powerful cores (~10)
  - Branch prediction, register renaming, etc.
- Large caches



CPU

#### • GPU

- Many number of cores (≥1000)
  - A lot simpler
- Small caches



Practically GPU can outperform CPU with parallelizable and relatively simple algorithms

Beomki Yeo

#### ACTS – A Common Tracking Software

- ACTS: C++ based track reconstruction toolkit for any experiment
  - General info about ACTS is covered in <u>A. Salzburger's talk</u> on Thursday
  - ATLAS Phase-II is the primary target (Offline & Online reconstruction)
- Most of development had happened with CPU
  - Also interested in implementing the CPU algorithms into hardware accelerators such as **GPU**
- Faced with following challenges during the initial attempts for GPU implementation:
  - Geometry and event data model designed with runtime polymorphism (*GPU-unfriendly*)
  - Difficult to recycle the original CPU algorithms for GPU
- Such challenges were the motivation for dedicated GPU R&D projects

#### ACTS GPU R&D Projects

#### • <u>traccc</u>

- Library for the GPU tracking demonstrator
- Candidate for ATLAS Online reconstruction

#### • <u>detray</u>

- Library for the GPU tracking geometry
- Compile-time polymorphism design
- There are also other cool R&D libraries (<u>VecMem</u>, <u>covfie</u> and <u>algebra-plugins</u>) but not covered in the today's presentation



### **Detray** – GPU Friendly Geometry

- Tracking geometry is very necessary for track finding and fitting
- **detray** moves out of the runtime polymorphism to adopt the compile-time polymorphism
- Index-based link between geometrical objects (No pointers!)
- Can translate ACTS geometry to utilize the existing plugins for other geometry libraries
  - o <u>DD4Hep</u>
  - <u>GeoModel</u> (ATLAS-specific)



#### **Detray** – Track Propagation

- Fourth order Runge-Kutta stepper inside the Inhomogeneous B field and material
- Material interaction: Bethe energy loss and multiple scattering
  - Non-gaussian noise (e.g. Brems) will be added in the near future
- Supports various surface types: Bound frame (e.g. pixels), Perigee frame (e.g. straw tubes and drift chamber)



Bound (left) and perigee (right) frame with a track intersecting

#### **Traccc** – Algorithms

- 1. Hit clusterization
  - Creating measurements from pixel readouts
- 2. Seeding (Pattern recognition)
  - Finding *three* measurements of a single particle
- 3. Track finding (Pattern recognition)
  - Finding all measurements of a single particle
  - Combinatorial Kalman Filter
- 4. Track fitting
  - Fitting the measurements to a track
  - Kalman Filter





2D Hit clusterization for measurement creation





Pattern recognition and fitting <u>Tracking algorithm in a nutshell</u>

#### Traccc – Workflow

- End-to-end analysis in the GPU
  - $\circ$   $\,$  Minimize the memory transfer between CPU and GPU  $\,$
- The tracking geometry is transferred to the GPU only once and the same geometry on the device is used for all events



#### Traccc – Development Status

- Mainly developed for CUDA (Nvidia exclusive) and SYCL (all vendors)
  - Ongoing development for portability solutions such as Alpaka, Kokkos and Futhark
- Supports both single (float) and double precisions

Category	Algorithms	CPU	CUDA	SYCL
Clusterization	CCL / FastSv / etc.	Image: A start of the start	~	
	Measurement creation			
Seeding	Spacepoint formation		Image: A start of the start	
	Spacepoint binning		Image: A start of the start	
	Seed finding		<ul> <li>Image: A set of the set of the</li></ul>	
	Track param estimation			
Track finding	Combinatorial KF			•
Track fitting	KF	<b>V</b>		<b>V</b>

#### Showcase: Open Data Detector

- Freely available open-source detector
  - LHC-like geometry
  - Developed by ACTS community for algorithm research and development
- Traccc team recently demonstrated full GPU tracking chain with the open data detector



#### Showcase: Open Data Detector (cont.)

 GPU throughput is higher than CPU for large size of data (pileup ≥ 140, corresponding to the ATLAS Run 4 data size)



HSF workshop, (May 16th 2024)

Beomki Yeo

### Showcase: ATLAS Experiment

- High Level Trigger for High Luminosity LHC
- Has been Investigated only for physical performance
  - Once GeoModel is fully supported, GPU speed performance can be measured Ο





May 18th 2024

Beomki Yeo

#### Showcase: CEPC

- Demonstrated seeding with the CEPC pixel geometry
- Initial performance looks good
  - See <u>Y. Zhang's talk</u> for detailed updates



Particle reconstruction efficiency of seeing algorithm with CEPC pixel geometry <u>ACTS Parallelization Meeting (Mar 8th 2024)</u>

#### Summary

- traccc is the GPU track reconstruction library as an R&D project of ACTS
- Recently showed promising results of **full GPU tracking chain** with the Open Data Detector
   Getting into the new R&D phase: performance optimization
- We welcome the participation of all experiments interested in GPU track reconstruction
   <u>ACTS Parallelization Meeting</u> (Biweekly Friday 16:00 CET / 22:00 Shanghai)

# Backups

### GPU Profiling Result for an Event

- Most of time is spent in GPU Kernels
- Main bottlenecks: track finding (CKF) and track fitting (KF) kernels
  - track propagation (e.g. matrix operations) is expensive



One event timeline (~0.1 sec per event)