### Data Handling in Allen at LHCb

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on behalf of the LHCb Real Time Analysis project

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# Triggering on heavy flavor decays





- Heavy flavor hadrons are long-lived and decay into low-momentum particles
- Can't effectively trigger on heavy flavor decays using hardware signatures
- Solution: process every event (30 MHz, 5 TB/s) in software

 $\overline{\text{Triggering}} \rightarrow \text{Real-Time Analysis}$ 

#### The Run 3 LHCb dataflow



Affordable computing power in a compact package



Courtesy of Dorothea Vom Bruch, arXiv:2003.11491

			global ar	id constar	nt memory	/		
	Block	(0,0) memory				Block	<b>k</b> (n, 0	
thread	thread	thread	thread		thread	thread	thr	
thread thread	thread	thread thread	thread thread		thread thread	thread thread	thr thr	

Block (n,0) shared memory						
$\mathbf{thread}$	thread	$\mathbf{thread}$	thread			
$\mathbf{thread}$	thread	thread	$\mathbf{thread}$			
thread	thread	thread	thread			

<b>Block</b> $(0,m)$							
shared memory							
_							
$\mathbf{thread}$	thread	thread	thread				
thread	thread	thread	thread				
thread	thread	thread	thread				

<b>Block</b> $(n,m)$						
shared memory						
thread	thread	thread	thread			
thread	thread	thread	thread			
thread	thread	thread	thread			

- Highly parallel processors with thousands of cores
- Relatively little memory
  - $O(100 \, \text{kB})$  "shared"
  - $\circ$   $\mathcal{O}(1\,\mathrm{GB})$  "global"
- **Same Instruction Multiple Data** computing model
- Many trigger and reconstruction tasks are parallelizable with some work!

	LHCb, Run 3	ALICE, Run 3	ATLAS/CMS, Run 4
Hardware trigger	No	No	Yes
Readout rate	$40\mathrm{MHz}\ pp$	$50\mathrm{kHz}$ PbPb	$\sim 1{\rm MHz}\ pp$
Data rate	$5\mathrm{TB/s}$	$3.5\mathrm{TB/s}$	$\sim 5\mathrm{TB/s}$

#### ALICE

- Large event sizes
- Reconstruction time dominated by TPC tracking with ~ 80× speedup on GPUs
- Clear use case for GPUs

#### LHCb

- Extremely high rate of small events
- No step dominates reconstruction time
- Advantages of GPUs compete with challenges

# The GPU technology decision (CSBS 6 (2022) 1, 1)



- Fully CPU- or GPU-based HLT1?
- GPU solution leads to cost savings on processors and networking
- Enough throughput headroom for additional features
- The final verdict: A GPU-based software trigger will allow LHCb to expand its physics reach in Run 3 and beyond

#### Allen is LHCb's GPU-based first level software trigger (HLT1)



- Decode data from the VELO, UT, SciFi, and Muon systems
- Cluster detector data into "hits"
- Build tracks (VELO, UT, and SciFi)
- Find primary vertices (PVs) (**VELO**)
- Match tracks to **Muon** hits

Work as a standalone application or as part of LHCb's software stack Can be compiled for CPU and GPU with CUDA or HIP

# Adapting to GPUs



Block $(0,m)$						
shared memory						
	thread	thread	thread	thread		
	thread	thread	thread	thread		
	thread	thread	thread	thread		

Block (n,m) shared memory					
threa	d thread	thread	thread		
threa	d thread	thread	thread		
threa	d thread	thread	thread		

#### Parallelization strategy

- Each block processes 1 event
- Use threads for intra-event parallelism

Take advantage of GPU computing power

- Some reconstruction tasks work very well as parallel algorithms
- Can be rewritten to expose parallelism

#### Deal with the GPU limitations

- Not all HLT1 tasks parallelize nicely
- Need to be rewritten to deal with memory limitations, optimize memory access patterns, etc.

# Example: VELO clustering





- 26 layers of silicon pixel detectors
- Identify seed candidates
- Cluster in constant time using bit masks

#### Example: VELO reconstruction



- Sort hits by azimuthal angle  $\phi$
- Create velo tracks
  - $\blacksquare$  Create triplets of hits at similar  $\phi$
  - Extrapolate track candidates to the next layer and look for hits
  - Repeat starting at the next layer

D. Campora, N. Neufeld, A. Riscos Núñez: "A fast local algorithm for track reconstruction on parallel architectures". IPDPSW 2019

#### Example: Redesigning the Allen Kalman filter



- Achieve the best possible estimate of track parameters at the state closest to the beamline
- Improve momentum resolution
- Best possible estimate of track impact parameter
- Uses detailed magnetic field map and detector description to propagate tracks
- Requires many expensive calculations

Can we design a Kalman filter to work with GPU constraints without sacrificing physics?

# Example: Redesigning the Allen Kalman filter



Parameterized Kalman filter CPC 265 (2021) 108026

- Parameterize track trajectories and multiple-scattering noise
- Calculations still expensive

Simplified Kalman filter

- Only fit the VELO segment
- $\blacksquare$  Two independent fits in x and y
- Produces close-to-optimal IP estimate
- No improvement in  $\sigma_p$  but this doesn't really matter in HLT1

Simplified fit contributes negligibly to Allen's execution time and memory footprint and accomplishes the physics goals of HLT1.

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# Making data GPU-friendly

Structure of Arrays (SOA) vs. Array of Structures (AOS)

- AOS:  $x_0, y_0, z_0, x_1, y_1, z_1, x_2, y_2, z_2$
- **SOA:**  $x_0, x_1, x_2, y_0, y_1, y_2, z_0, z_1, z_2$

No dynamic memory allocation during kernel execution

- No push\_back or emplace\_back!
- Need to allocate global memory before kernel execution

GPU memory accesses are expensive

- Copying an entire structure from an AOS may take many global memory accesses
- Can often coalesce memory accesses using SOA formats

Allen data is generally stored in simple C arrays with corresponding arrays of offsets that provide structure.

```
unsigned offset = offsets[event_number];
unsigned n_tracks = offsets[event_number + 1] - offset;
for (unsigned i_track = 0; i_track < n_tracks; i_track++) {
  float px = track_data[offset + i_track];
  float py = track_data[offset + n_tracks + i_track];
  float pz = track_data[offset + 2*n_tracks + i_track];
...
}
```

#### $\mathbf{Vs.}$

```
for (auto track : tracks) {
  float px = track.px();
  float py = track.py();
  float pz = track.pz();
   ...
}
```

- Keeping track of arrays and offsets becomes difficult for complex data structures
- Changes to underlying data structures require changes to downstream code
- Solution: users access slices of arrays using views
- Handles indices and offsets behind the scenes

# Not just GPUs...

#### CPU HLT1 throughput



Designing algorithms and data structures to take advantage of parallelization is going to become more and more important regardless of hardware choice!

# Computing performance



• Can handle the full LHC event rate with  $\sim 200$  GPUs

• Throughput scales well with theoretical computing power

# Physics performance



- Huge improvement over hardware level triggers
- Accomplishes the physics goals defined in the Upgrade Trigger TDR
- Provides throughput headroom for additional algorithms, like ECAL decoding and electron ID

- Using GPUs will allow LHCb to expand its physics program during Run 3 and beyond
- Taking advantage of GPUs requires redesigning algorithms and data structures
- These design considerations will benefit both GPU- and CPU-based systems

# Thank You!