RICH Status

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Introduction

- Focus on reconstruction developments.
 - Work still ongoing in the simulation, but not covered in detail here.
- RICH in the upgrade has :-
 - **Upgraded photon detectors** (MaPMTs replacing existing HPDs)
 - Updated RICH1 optics (re-optimisation taking into account Aerogel removal).
- Fundamentally though, works in the same way. Same algorithms (currently) envisaged.
 - Have had an 'upgrade' sequence running based on update detector description for a long while. Formed basis of MC design work and TDR results.
- Present here recent developments using the upgrade Gaudi Framework.
 - Migration to the 'Functional' framework
 - CPU improvements to the RICH photon reconstruction
 - First push towards fully utilising SIMD vectorisation
 - Ready now and being deployed for 2017 data taking.

RICH Reconstruction Basics

RICH Detector Hits

- Raw Event Decoding (HPD/MaPMT)
- Pixel Clustering
- Pixel Space Points (global and local)

Cherenkov Photon Candidates

- Reconstruction from pixels and segments
- Predicted likelihood contribution to each pixel

Tracks (Long, Downstream, Upstream)

- Radiator trajectories (segments)
- Ray traced PD panel impact points (global and local frames)
- Cherenkov photon expected yields
 and energy spectra
- Ray traced Cherenkov cones
- Geometrical efficiency
- Expected Cherenkov Angles
- Expected Cherenkov resolutions

Global Likelihood Minimisation

- Compute pixel backgrounds
- Minimise global event likelihood. Computes the Rich 'DLL' values.

Original Software Design

- Current C++ reconstruction dates back at least 10 years.
- Designed with very different ideas in mind to now
 - Object oriented design was the next big thing.
 - Parallel execution never entered the picture.
 - Basic *SIMD instructions* may have been available, but never discussed.
- Different constraints from the framework.
 - Originally *no python configuration*. Flat 'job options' files. Very limited.
- Lead to a software model which has served well, but is now showing its age.
- Two main parts to RICH design :-
 - Reconstruction event model. Tracks, Segments, Pixels and Photons.
 - **Data provider tools**. Event model object cache data and tool compute and fill ondemand ('lazy' evaluation model).

"RichRecEvent"



* Only {0,1,2} since Run2 due to Aerogel removal

Data Cache Example - 'RichRecPhoton'

Private Attributes

Rich::ObjPtn< LHCb::RichGeomPhoton >	m_geomPhoton The result of the Cherenkov angles reconstruction. More
Rich::HypoData< LHCb::RichRecPhoton::FloatType >	m_expPixelSignalPhots The Expected number of signal photons for each particle hypothesis. More
Rich::HypoData< LHCb::RichRecPhoton::FloatType >	m_expPixelScattPhots The Expected number of scattered photons for each particle hypothesis. More
LHCb::RichRecSegment *	m_richRecSegment RichRecSegment used in the formation of the RichRecPhoton candidate. More
LHCb::RichRecTrack *	m_richRecTrack RichRecTrack associated to the RichRecSegment used in the formation of the RichRecPhoton candidate. More
LHCb::RichRecPixel *	m_richRecPixel RichRecPixel used in the formation of the RichRecPhoton candidate. More

- <u>m_geomPhoton</u> : Encapsulates the photon reconstruction
- <u>m expPixel{Signal,Scatt}Phots</u> : Cached data on expected detector signals.
- <u>m_richRec{Segment,Pixel,Track}</u> : Relations back to associated data objects
- Data only related to each other in that they all are associated to the same photon candidates.
- RichRecPhoton a small cache ! Segment class much more extensive.

Moving to the Future...

- Make the algorithms fully thread safe.
- Address design choices that hinder modern compiler optimisations.
 - Algorithms that operate on const vector<Foo>& rather than tools on Foo*. Create vector<Baa>.
 - Move towards SOA rather than AOS designs (so small Foo and Baa objects, often just float).
 - Remove monolithic event data model 'cache' objects. More, smaller, more focused data objects. Composition when required using Range v3 library.
- Make use of AnyDataHandle to remove DataObject inheritance constraint.
- Move processing to algorithms. Use tools only where appropriate.
- Remove all monitoring (even basic tallies) from reconstruction algorithms. Keep streamlined.
- Simplify python configuration.
- Lower barrier for new people to contribute
- · <u>Conclusion Start from scratch.</u>

Status of Functional Migration

- Fully functional RICH reconstruction sequence using the new framework available in latest Brunel release.
 - In fact, is the default processing sequence for this years data taking.
 - Old sequence still available, as 'backup'.
- Timing wise, *new sequence is ~2.5 times faster** than that in the Brunel released for the start of the 2016 data taking.
 - Largely due to the adoption of guidelines to better format the data how the CPU wants it, not the user. e.g. utilise SIMD vectorisation.
 - Also in part due to some specific code improvements.

* Comparing Brunel v50r1 to v52r2, using same machine and 1000 event sample

SIMD (Single Instruction Multiple Data)

- Hardware vendors raise computational power of today's CPUs with increasing support for parallelism:
 - More cores (beyond the scope of this talk)
 - Larger vector units, richer vector instruction sets
- Vector units: perform same operation on multiple data
 - Data parallelism at instruction level
- Peak performance achievable only if vector units are properly used
 - Especially for "extreme" architectures like the Xeon Phi



https://indico.cern.ch/event/258092/contributions/1588517/attachments/454175/629578/ACAT2014_ModernisingROOT_SWenzel.pdf

Horizontal (external) vectorisation:



Vertical (internal) vectorisation:

Object
{ X,
У,
Z
}

vectorize internally the algorithm operating on a single object

- example: for loop over internal data
- Object data member (e.g. x,y,z) must be stored in a vector

https://indico.cern.ch/event/258092/contributions/1588517/attachments/454175/629578/ACAT2014_ModernisingROOT_SWenzel.pdf

Ray Tracing Photons



- Ray tracing photons through RICH mirror system to detector plane major CPU usage for the RICH
- Simple geometrical calculation repeated many times. Should be easy to vectorise.
- Uses ROOT GenVector library. (The ubiquitous Gaudi::XYZVector and Gaudi::XYZPoint types.)
- Have previously tried internally vectorising (vertical) using libraries like Eigen. Results not too impressive. Difficult to see how this approach can fully utilise SIMD capabilities.
- Look instead at using Vc library. Provides Vc::float_v, Vc::double_v types that behave (more or less) like float, double, but are vectors(arrays) of N values (N depends on SIMD level). Horizontal vectorisation.

Reflect Spherical - Scalar

```
template < typename POINT, typename VECTOR, typename FTYPE >
inline
typename std::enable if< std::is arithmetic<typename POINT::Scalar>::value &&
                         std::is arithmetic<typename VECTOR::Scalar>::value &&
                         std::is arithmetic<FTYPE>::value, bool >::type
reflectSpherical ( POINT& position,
                   VECTOR& direction,
                   const POINT& CoC.
                   const FTYPE radius )
  constexpr FTYPE zero(0), two(2.0), four(4.0), half(0.5);
  const FTYPE
                   a = direction.Mag2();
  const VECTOR delta = position - CoC;
  const FTYPE
                  b = two * direction.Dot( delta );
  const FTYPE
                   c = delta.Mag2() - radius*radius;
  const FTYPE discr = b*b - four*a*c;
                 OK = discr > zero;
  const bool
 if ( OK )
    const FTYPE dist = half * ( std::sqrt(discr) - b ) / a;
   // change position to the intersection point
    position += dist * direction;
    // reflect the vector
    // r = u - 2(u.n)n, r=reflection, u=incident, n=normal
    const VECTOR normal = position - CoC;
    direction -= ( two * normal.Dot(direction) / normal.Mag2() ) * normal;
  return OK;
```

 POINT and VECTOR are just template types for, e.g. Scalar XYZPoint<double> and XYZVector<double>

Reflect Spherical - Vectorised

```
template < typename POINT, typename VECTOR, typename FTYPE >
inline
typename std::enable if< !std::is arithmetic<typename POINT::Scalar>::value &&
                         !std::is_arithmetic<typename_VECTOR::Scalar>::value_&&
                         !std::is_arithmetic<FTYPE>::value,
                         typename FTYPE::mask type >::type
reflectSpherical ( POINT& position,
                   VECTOR& direction,
                   const POINT& CoC,
                   const FTYPE radius )
  const FTYPE two(2.0), four(4.0), half(0.5);
                  a = direction.Mag2();
  const FTYPE
  const VECTOR delta = position - CoC;
  const FTYPE
                  b = two * direction.Dot( delta );
  const FTYPE
                   c = delta.Maq2() - radius*radius;
              discr = b*b - four*a*c;
  FTYPE
  typename FTYPE::mask type OK = discr > FTYPE::Zero();
  if ( any of (OK) )
    // Zero out the negative values in discr, to prevent sqrt(-ve)
    discr(!OK) = FTYPE::Zero();
    // compute the distance
    const FTYPE dist = half * ( sqrt(discr) - b ) / a;
    // change position to the intersection point
   position += dist * direction;
    // reflect the vector
    // r = u - 2(u.n)n, r=reflection, u=incident, n=normal
    const VECTOR normal = position - CoC;
    direction -= ( two * normal.Dot(direction) / normal.Maq2() ) * normal;
  // return the mask indicating which results should be used
  return OK;
```

- POINT and VECTOR are *vectorised* e.g., XYZPoint<Vc::double_v> and XYZVector<Vc::double_v>
- Effectively, think of each as being an 'array' of Vc::double_V::Size entries, operated on in parallel

Vectorised Ray Tracing - Performance

 Idealised test application - photon reflection of first primary (spherical) and then secondary (flat) mirrors - shows almost 'perfect SIMD' speed up.



Creating 9600 random photons ... ROOT double 228752 Eigen double 242017 speedup 0.94519 Vc double 58243 speedup 3.92754 ROOT float 209756 Eigen float 212881 speedup 0.98532 Vc float 26539 speedup 7.90369

- Real world deployment, somewhat more complicated...
 - Multiple mirror segments, HPD/MaPMT intersection. Not fully vectorised yet.
 - Nevertheless, already see <u>~33% speed up</u>.
- First steps... Will significantly improve once we learn how to work with SIMD vectorisation better.

Ongoing 'Parallel' Efforts

- Studies ongoing to port expensive RICH tasks (photon reconstruction, ray tracing) to various 'highly parallel' architectures (R.Schwemmer, C.Quast, C.Faerber)
 - <u>KL (Knights Landing)</u> 'Regular' x86 CPU with many cores and SIMD vector units.
 - FPGAs. Very (very) parallel...
- Work ongoing, but results already encouraging and have resulted in some improvements which could be incorporated in the last release.
 - Speed up in the quartic reconstruction algorithm⁴ due to replacement of an analytic solution with a *Newton Raphson approach*.
 - O(40%) speed up !
- <u>Clear example of the advantage of getting new people</u> <u>involved.</u>





Performance statistically identical

Conclusions

- RICH Reconstruction ported to the upgrade 'functional' framework.
 - Deployed and ready for 2017 data taking.
 - Directly, and indirectly, has resulted in a significant improvement in CPU time.
 - A factor of 2.5 w.r.t. 2016 reconstruction.
 - RICH now a rather small fraction of overall reconstruction time.
- Just the start...
 - More improvements expected through further improved utilisation of CPU/ cache friendly data structures.
- So far have only touched 'internal' RICH data structures. Eventually the **public event model** (RichPID, tracks, Particles) will need to be heavily updated.