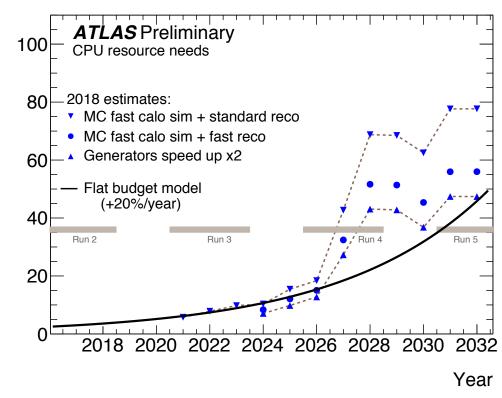


Acts in multi-threaded environment

Paul Gessinger - 01/15/2019 - Tracking workshop for HEP - LBNL

What's the problem, where are we?

- Track reconstruction is (often) the most CPU-intensive part of event reconstruction (ATLAS: ≈80%)
- CPU resources are limited!
- Scaling with pile-up is not encouraging
- Parallelization can help!
- But: just executing on more CPU(-corest might be problematic: $\frac{memory}{\#CPUs}$ will decrease
- Other approaches can help to saturate more CPUs on same amount of memory



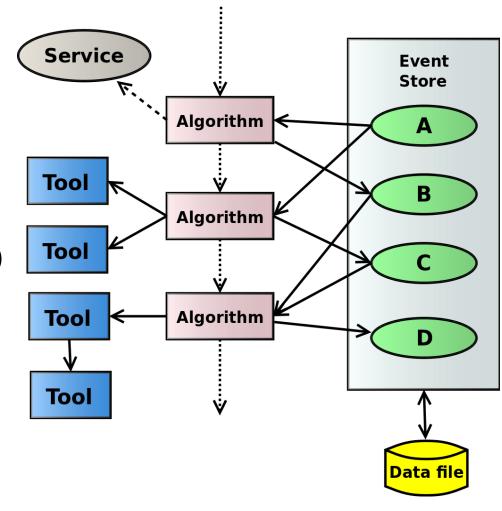
Annual CPU Consu

Parallelization in ATLAS

ATLAS reconstruction

- ATLAS reconstruction comprises multiple domains, (e.g. ID, muons, calorimeters, jets) with lots of interdependencies
- Time spent by domain:
 - ID: (Pixel, SCT, TRT) ≈ 11s/event
 - Everything else: $\approx 15 \times 1.5$ s (15 = no. of domains)
- ID reconstruction is clearly the place to optimize!
- ATLAS software is based on Gaudi
 - Algorithms process events, data flows through input and output collections
 - Can use Tools to offload some of their work
 - Services (singletons) can be accessed from both

Sequential processing



[1]

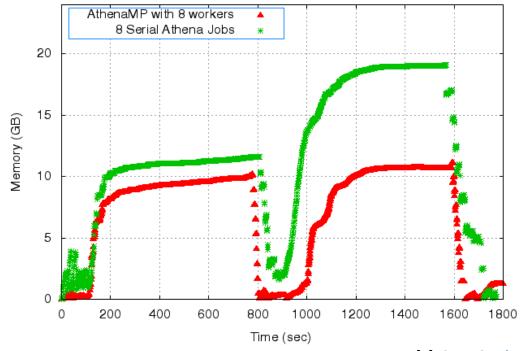
Run separate jobs

- Simplest idea: just run multiple instances of the whole software
- Almost "trivial" to implement
- But:
 - Can only parallelize on event ranges
 - Duplicates the entire software stack in memory!

Multi processing

- Run 2: parallelization with multiple processes, but forked
- Requires little change to actual code
- Start processing first event, then fork the process to all CPUs
- Copy-on-write allows easy sharing of memory
- Memory savings might not be enough for Run 3 and beyond

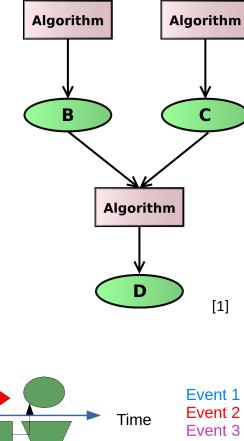


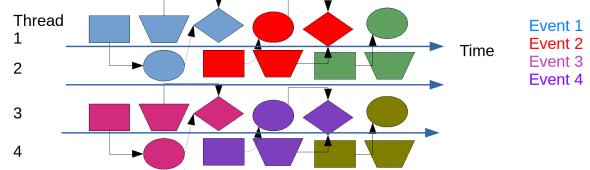


[1]: Scott Snyder

Multi-threading

- AthenaMT event-processing framework (in development)
- Allow parallelization at algorithm level (scheduler figures out data dependencies)
 - Inter- and intra-event parallelism possible
- Ideally: algorithms only instantiated once, invoked for every event
 - Keeps memory footprint low
 - All tools need to be thread-safe
 - Most importantly: no mutable state
- However: most of ID chain is sequential





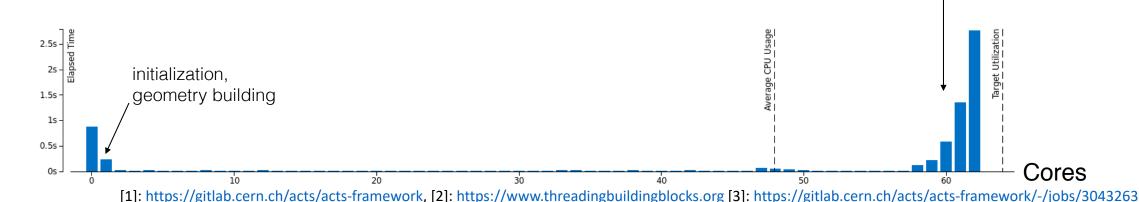
[1]: Scott Snyder

Enter: Acts

Toolkit containing thread-safe algorithms and utilities

Acts and Multi-Threading

- Acts is designed to be thread-safe:
 - Local configuration and state
 - Immutable data everywhere (e. g. geometry)
- Acts does not provide infrastructure for parallelization
- Parallelization is implemented by experiment software
- The test framework¹ contains a TBB² based event-by-event parallel loop
 - CI job³ tests results from demo particle extrapolation is identical, for $n_{threads}=1$ and $n_{threads}>1$



Design choices for multi-threading

Mutable state is limited to thread-local storage, passed around explicitly
 OptionsType eOptions;

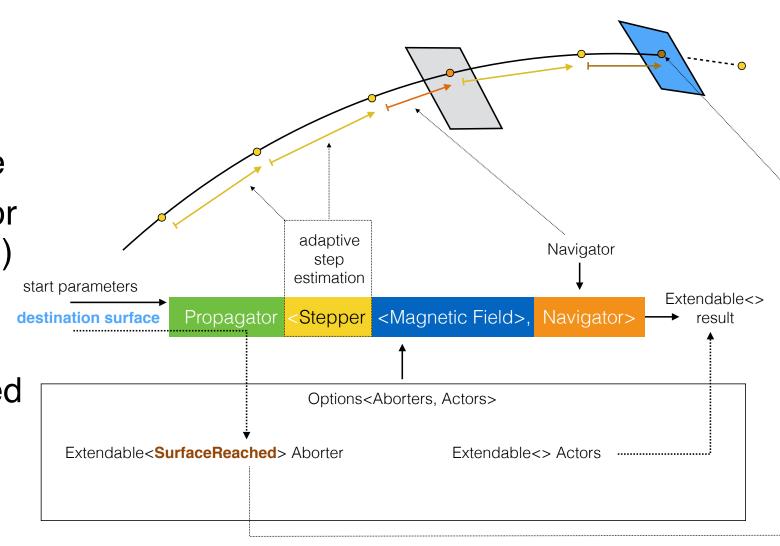
```
OptionsType eOptions;
// setup options ...
StateType state(start, eOptions);

// Perform the actual propagation & check its outcome
if (propagate_impl(result, state) != Status::IN_PROGRESS) {
   result.status = Status::FAILURE;
} else {
   // Convert into return type and fill the result object
   m_stepper.convert(state.stepping, result);
   result.status = Status::SUCCESS;
}
return result;
```

 Geometry is considered immutable after creation (technically, we const-cast it currently during closure, but we want to get rid of that)

Implementation Propagator and KalmanFilter

- Acts propagation is based on existing ATLAS implementation
- Redesigned interface, eliminated all mutable state
- Switched to Eigen library for math (much more readable)
- Designed to be flexible and extensible!
- KalmanFilter is implemented on top of the Propagator



Implementation Propagator and KalmanFilter

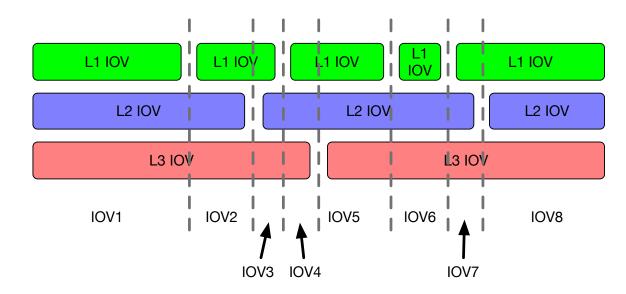
- KalmanFilter contains Propagator object
- Call to fit() sets up propagation and attaches the KalmanActor
- KalmanActor implements logic:
 - Initialize
 - Forward filter
 - Backward smoothing
- Delegates to separate calibrator, updator and smoother
- Can be set up once, then invoked from many threads

```
template <typename input measurements t,
         typename parameters t,
         typename surface t>
auto
fit(input measurements t measurements,
   const parameters t& sParameters,
   const surface t*
                        rSurface = nullptr) const
 // Bring the measurements into Acts style
 auto trackStates = m inputConverter(measurements);
 // Create the ActionList and AbortList
 using KalmanActor = Actor<decltype(trackStates)>;
 using KalmanResult = typename KalmanActor::result type;
 using Actors
                     = ActionList<KalmanActor>;
 using Aborters
                     = AbortList<>;
  // Create relevant options for the propagation options
  PropagatorOptions<Actors, Aborters> kalmanOptions;
  // Catch the actor and set the measurements
 auto& kalmanActor = kalmanOptions.actionList.template get<KalmanActor>();
 kalmanActor.trackStates = std::move(trackStates);
 kalmanActor.targetSurface = rSurface;
  // Run the fitter
  const auto& result
     = m propagator.template propagate(sParameters, kalmanOptions);
  /// Get the result of the fit
 auto kalmanResult = result.template get<KalmanResult>();
  // Return the converted Track
 return m outputConverter(std::move(kalmanResult));
```

Challenges and problems

Conditions and concurrency

- Conditions: parameters recorded during data taking
- Vary between events (Interval of Validity, IOV)
 - Time-dependent detector properties (wire-sagging, bending, temperatures, ...)
 - E.g. alignment of sensitive surfaces can change over time (fitted empirically to account for it)
- Problem: we consider the geometry constant!
- One solution would be: re-create the geometry at IOV boundaries

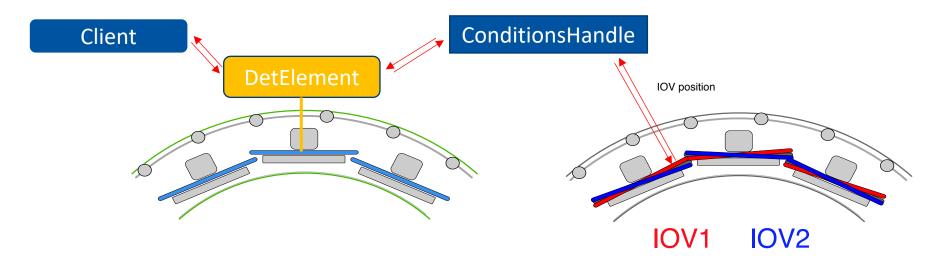


Conditions and concurrency

- Previously: sequential event processing
 - Conditions are accessed through services
 - When event processing reaches an IOV boundary, all services are notified
 - Services update their conditions caches
 - Event processing continues
- Problem: this does not work with multiple threads: multiple IOVs can be in flight at the same time
 - n_{IOV} might be small for unfiltered data, but could be large for highly selected samples
- Solution in ATLAS: remove conditions services, enter: Conditions algorithms
 - Conditions algorithm makes conditions data available
 - Scheduler makes sure they are scheduled before conditions are required.
- Problem: we would potentially re-build the geometry quite a lot!

Conditions: alignment

- What information do we need for the alignment? Module positions (transforms)
- Tracking geometry consists of surfaces, are connected to a (experiment specific) detector element
- Transform comes from that detector element
- Detector elements need to be aware of alignment / IOV!



Conditions: alignment

- However: multiple alignments can be in flight at the same time!
 - Detector elements needs to know from which event they are accessed
- ATLAS: Conditions service provides conditions objects for an event context
 - Algorithms are explicitly passed the event context:
 StatusCode
 ActsExtrapolationAlg::execute_r(const EventContext& ctx)
 const {

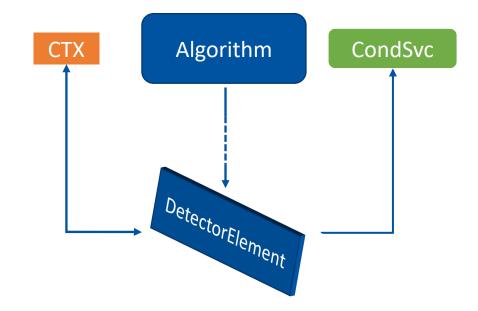
• **Problem**: call chain from algorithm down to individual surfaces is **very** deep and has many paths.

// ...

Conditions: alignment

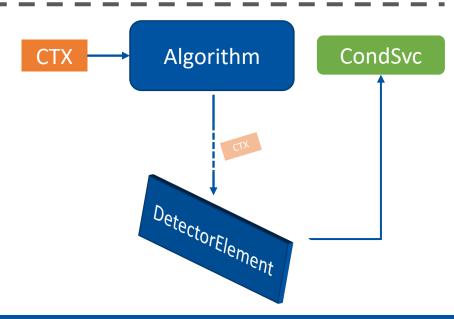
• The magic solution:

- Event context can be accessed from threadlocal static variable (set up by the scheduler)
- Detector element can implicitly figure out which IOV it is being accessed from!
- (this is what was implemented so far)



• The clean solution:

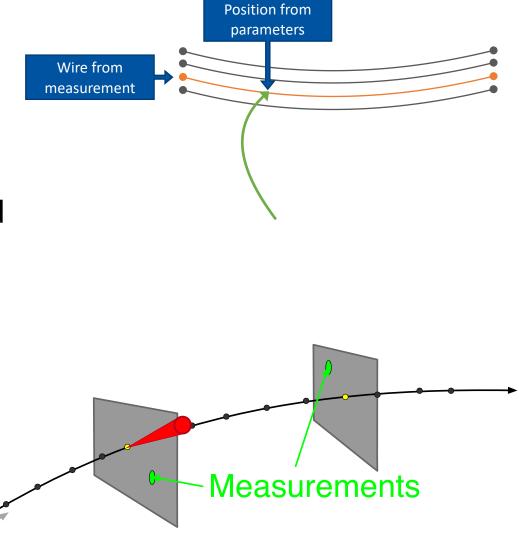
- Every call chain down to sensitive surfaces needs to pass along the context
- Type-erase context object so it can be experiment-agnostic
- (we will most likely switch to this solution)



Conditions for calibration

- Calibration is performed during fitting
 - Idea: raw measurement supplemented with full track parameter prediction
 - Typical example: wire-sagging: wire identified by measurement, sagging calculated at predicted position
- The updator delegates to the calibrator

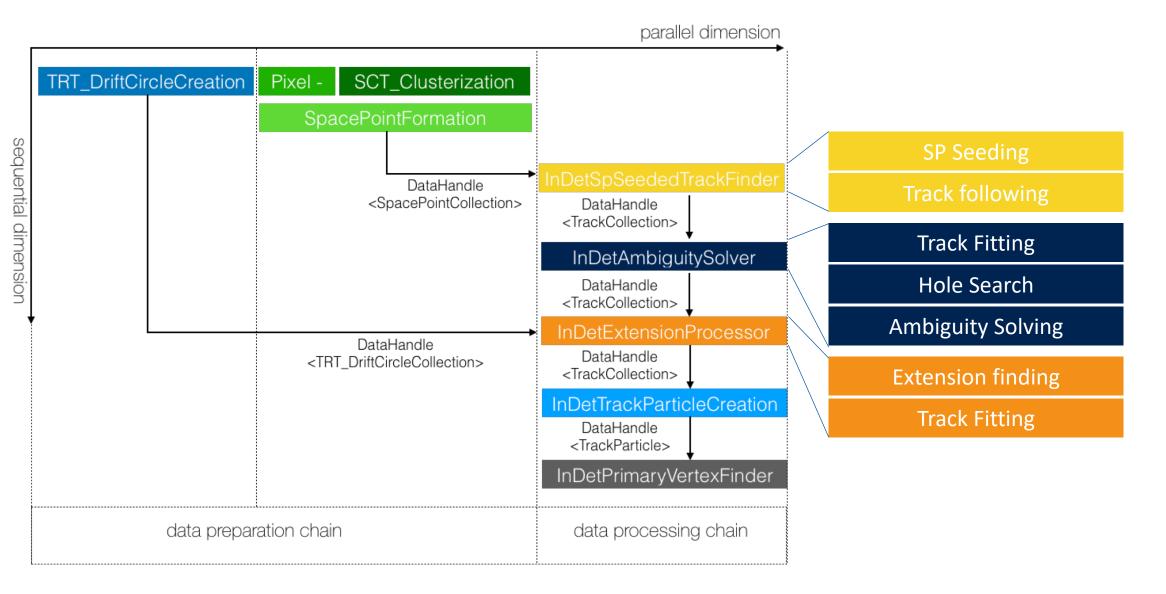
 Calibrator needs access to conditions (not a problem, since experiment specific)



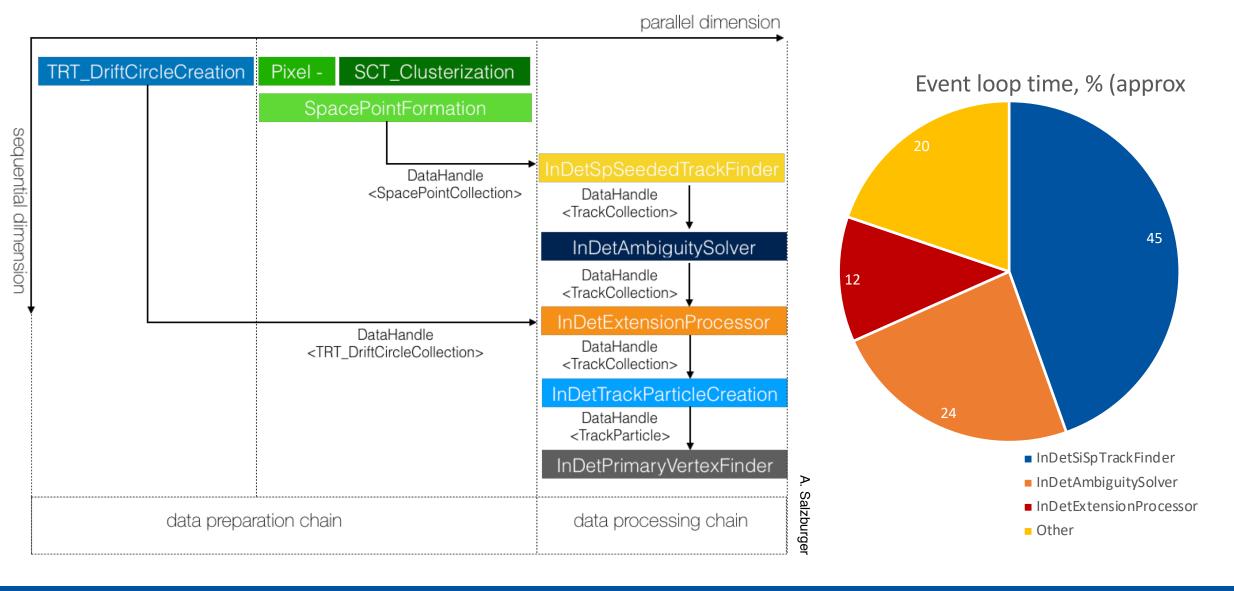
Opportunities

From ATLAS' perspective

Algorithms in ID reconstruction



Time spent in ATLAS reconstruction



Algorithm	Concurrent Cell	Number of cells	Tests	Rel. CPU of ID	Comments
Cluster creation	per module	O(1000)	no tests exist	O(5%)	output merging step necessary
SpacePoint creation	per space point	O(10000)	no tests exist	O(<1%)	output merging step necessary
SpacePoint seeded track finding	detector region	O(10-100)	GPU based test version from 2011	O(50%)	overlaps are dangerous
Ambiguity Solving	fitting: per track ambiguity: per tracks through shared modules	O(100-1000)	no tests exist	O(20%)	book keeping of hits is shared
TRT extension	per track	O(100)	no tests exist	O(10%)	may need update for DataHandle

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Parallelization of ID reco chain

- The largest parts of the ID reco chain are sequential
 - InDetSpSeededTrackFinder needs to run before Ambiguity solving, and so on
- Inputs (drift circle creation, clusterization) can be parallelized
 - Need to be careful not to have runaway preemptive pre-processing and starve main algorithm
- If the large algorithms can be made thread-safe: event-by-event parallelism possible
 - If CPUs can be saturate like this: fine, we're done!
 - If not? Finer granularity!
- In many cases: parallelism by geometric regions requires careful treatment of boundaries (ambiguity solving)

Parallelization of ID reco chain

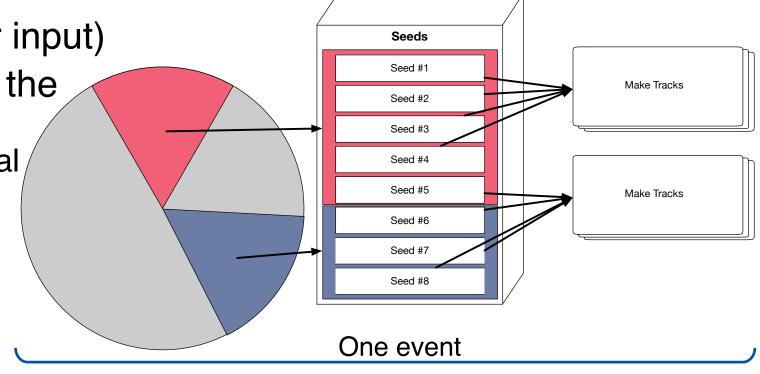
- E.g. produce seeds parallelized, batch and spawn tasks to do track following
- Acts seed finder implementation can parallelize on middle space point bins!

 Feed into track finding (which runs sequential per input)

 Tune chunk sizes to strike the right balance

 No effort so far beyond initial tests

 Should allow scaling parallelism as necessary



Summary

- Multi-threaded infrastructure is well underway in ATLAS
- Acts provides components that can be deployed in concurrent environments
- Details of unit-of-parallelization is left to the experiment to decide