

Track reconstruction and beyond

Xiaocong Ai Dec 6, 2021

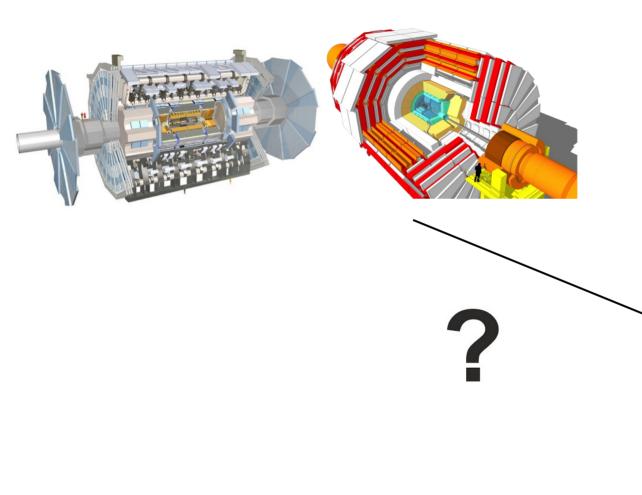




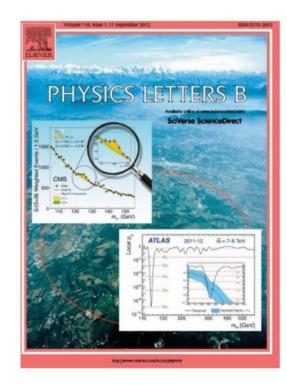
Outline

- Event reconstruction at HEP experiments
- Tracking is pivotal
- Tracking strategies
- The tracking challenges
- How to achieve accurate, efficient and fast tracking for various detectors
- Track-based detector alignment
- Summary

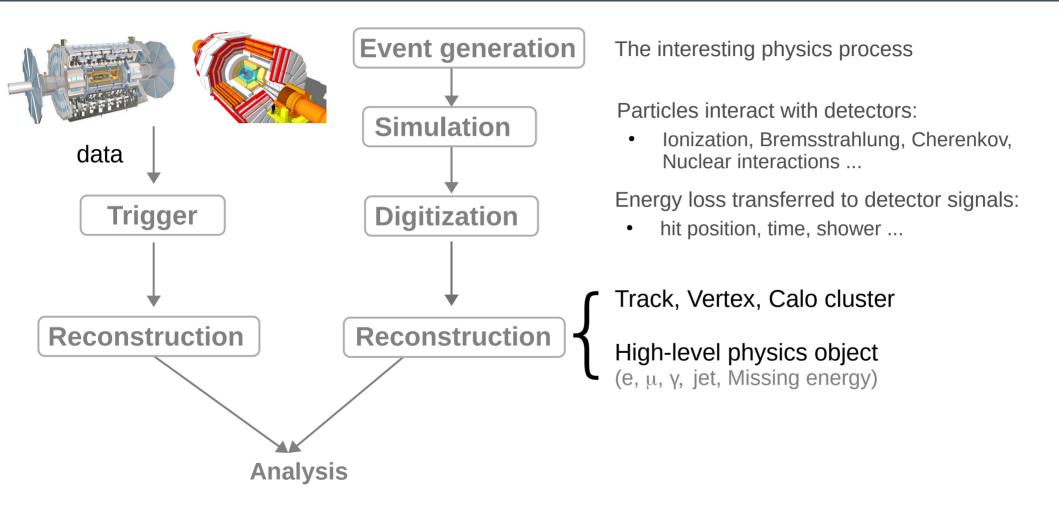
The long long story...



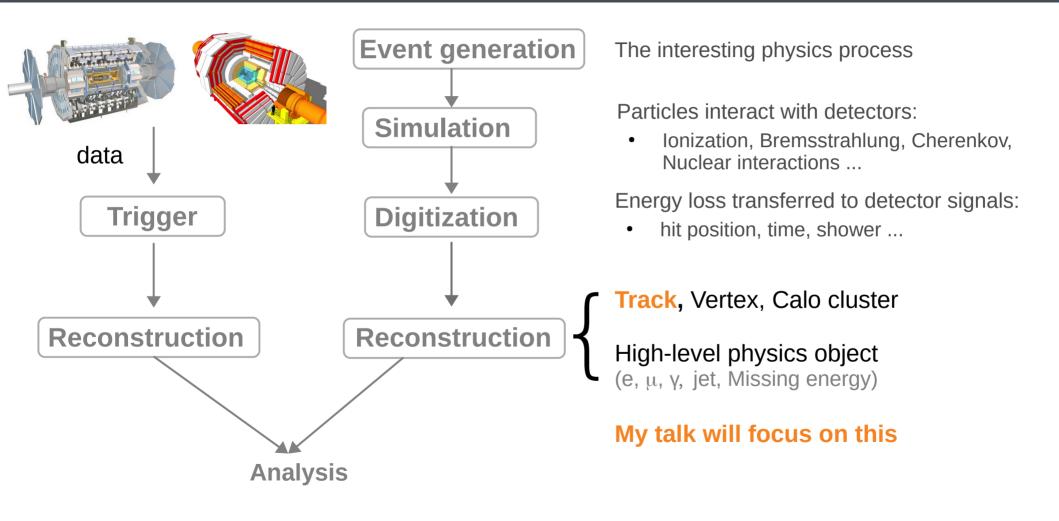
The exciting papers



Event reconstruction at HEP experiments



Event reconstruction at HEP experiments



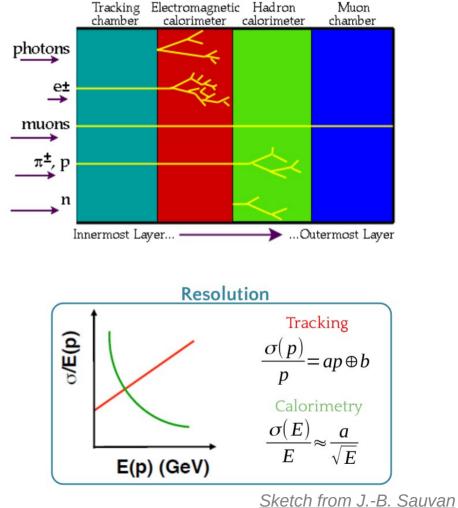
Tracking is pivotal

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What is tracking

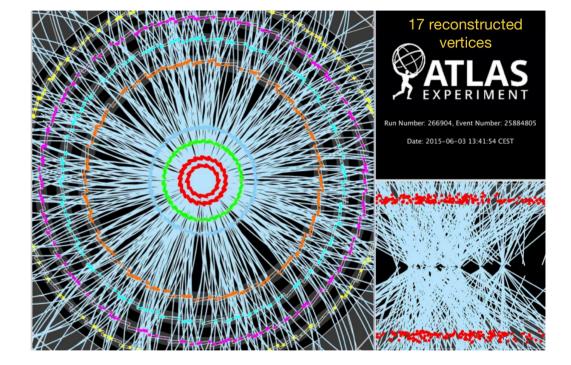
- Reconstruct charged particle (e, μ, charged hadrons) trajectory from tracker signals
- Estimate charged particles properties
 - Momentum via curvature in B field
 - Charge
 - Origin and direction
 - Velocity (dE/dx)

Tracker, calorimeter and muon chamber are complementary to each other!



Tracking is about vertex reconstruction

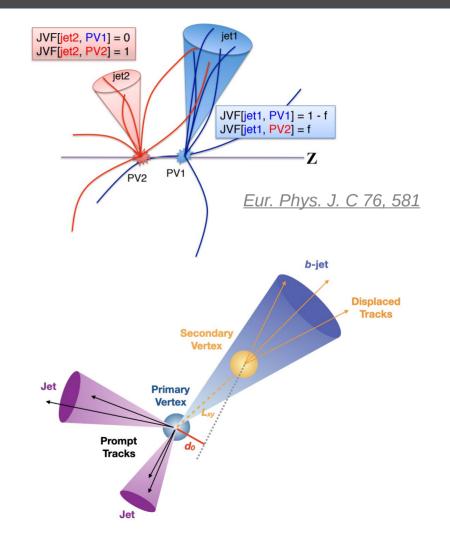
- Primary vertex reconstruction use estimated track parameters of charged particles as inputs
 - Vertex finding
 - Associate tracks to vertices
 - Vertex fitting
 - Estimate vertex position



Tens of additional proton–proton collisions accompanying the hard-scatter interaction, i.e. pile-up (μ)

Tracks/vertices are not just about charged particles

- Jets and missing energy reconstruction
 - Better pT resolution for low pT tracks and angular resolution provided by tracker
 - Tracks/vertices are crucial for pile-up mitigation (needs precise jet-vertex association)
- Jet flavor-tagging (b, c or light-flavor jet)
 - Impact parameters, secondary vertices and length of flight
- And track-based alignment of detectors!



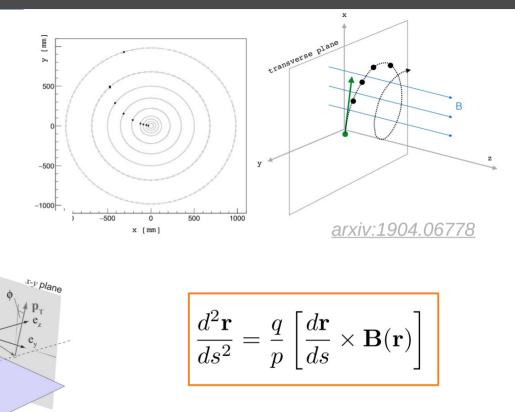
Tracking strategies

Track parameterization

- Helix trajectory of charged particle in homogeneous solonoid magnetic field
- Described by five (or six) parameters

x-y plane

- e.g. L = (loc0, loc1, phi, theta, q/p, t)



Solved numerically using <u>Runge-Kutta-Nyström method</u>

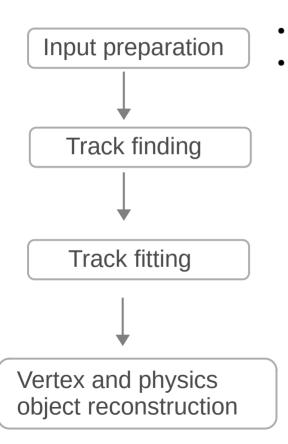


track

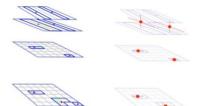
Track parameter represented at detector local surface

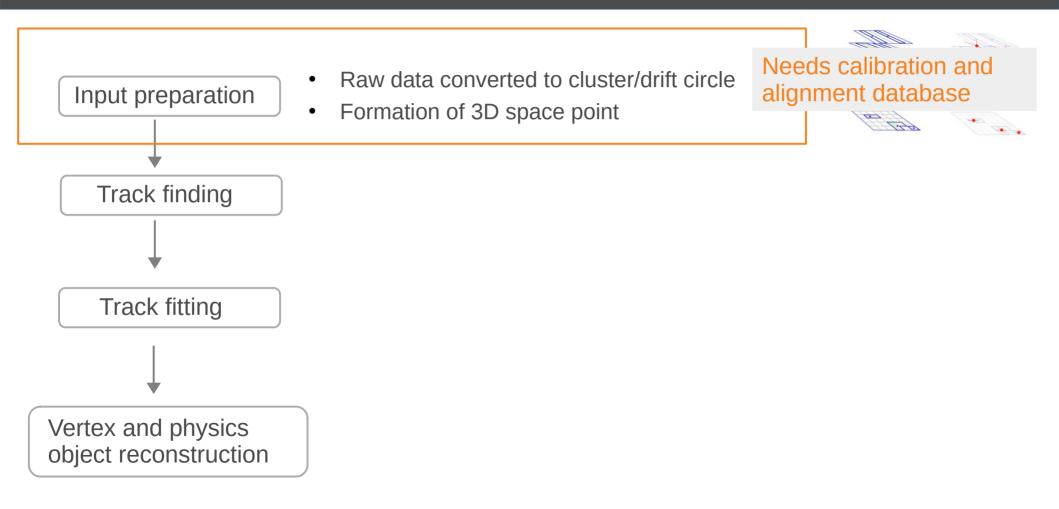
From E. Moyse

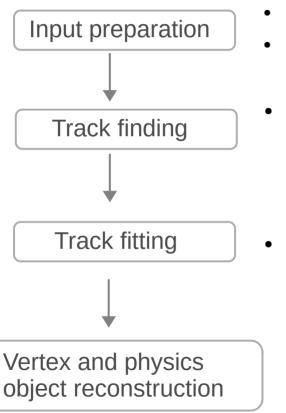
track



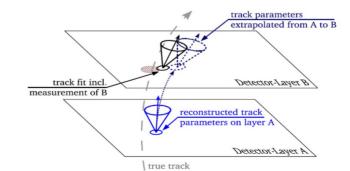
- Raw data converted to cluster/drift circle
- Formation of 3D space point





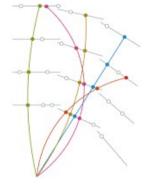


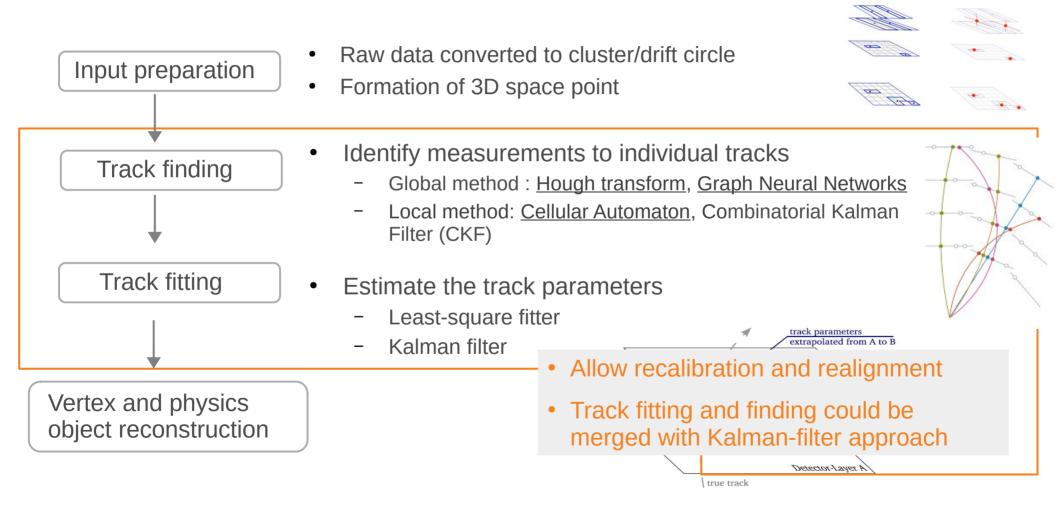
- Raw data converted to cluster/drift circle
- Formation of 3D space point
- Identify measurements to individual tracks
 - Global method : Hough transform, Graph Neural Networks
 - Local method: <u>Cellular Automaton</u>, Combinatorial Kalman Filter (CKF)
- Estimate the track parameters
 - Least-square fitter
 - Kalman filter



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The Least-square fitter for track fitting

- Simultaneously taking into account all measurements
- Cons:
 - Computationally expensive (large size matrix operation,
 - Consideration of material effects is non-trivial
 - Extensions for non-Gaussian noise and non-linear models are difficult

$$\mathcal{P}_i(m_i, \boldsymbol{\lambda}) = \frac{1}{\sqrt{2\pi}} \exp\left[\frac{1}{2} \left(\frac{m_i - h_i(\boldsymbol{\lambda})}{\sigma_i}\right)^2\right]$$

$$\chi^2 = \sum_i \left(\frac{m_i - h_i(\boldsymbol{\lambda})}{\sigma_i}\right)^2 = -2\ln\mathcal{L} + \text{const.}$$

The Least-square fitter for track fitting

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 - Computationally expensive (large size matrix operation) $\chi^2 = \sum_i \left(\frac{m_i h_i(\lambda)}{\sigma_i}\right)^2 = -2 \ln \mathcal{L} + \text{const.}$
 - Consideration of material effects is non-trivial
 - Extensions for non-Gaussian noise and non-linear models are difficult
- Superseded by the Kalman filter, but still used if at least material effects can be considered, e.g. the <u>General Broken Lines (GBL)</u> χ^2 fitter
 - Used for e.g. **beam test tracking** for detector characterization

 $\mathcal{P}_i(m_i, \boldsymbol{\lambda}) = \frac{1}{\sqrt{2\pi}} \exp\left[\frac{1}{2} \left(\frac{m_i - h_i(\boldsymbol{\lambda})}{\sigma_i}\right)^2\right]$

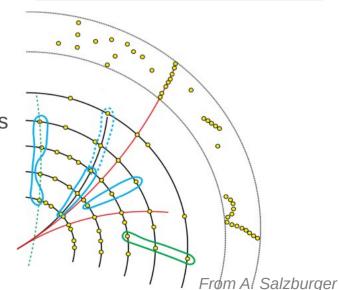
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x^{trk}-x^{nit}[mm]

The Kalman Filter-based track fitting and finding

- Progressively consider measurements
 - Extrapolate from k-1 to k iteratively: prediction + filtering
 - Backward smoothing when forward filtering is done
- Pros:
 - Straightforward handling of material effects
 - Allow track finding alongside fitting using CKF
 - Extension-friendly
 - For non-gaussian noise, e.g. bremsstrahlung energy loss
 - → Gaussian Sum Filter (GSF)
 - For non-linear measurement model
 - \rightarrow Second-order KF!

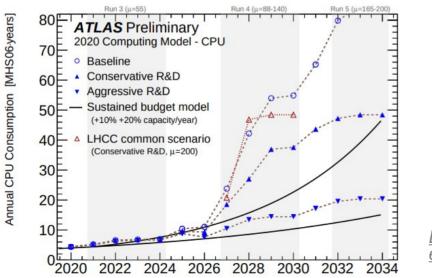
<u>Comput Softw Big Sci 5, 20 (2021)</u>



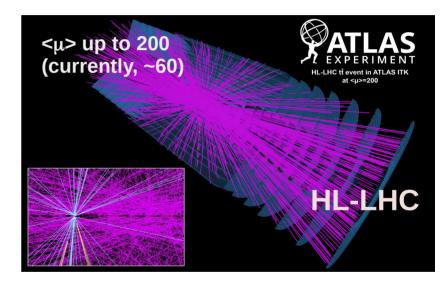
The tracking challenges

Much more dense environment

- Much increased combinatorics with high pileup at future hadron colliders, e.g.
 - ~7k particles/event with <μ> = 200 at High Luminosity LHC (HL-LHC)
 - $<\mu> = 1000$ at FCC-hh
- Much increased CPU needs



More sensitive to rare phyics, and far more combinatorics!

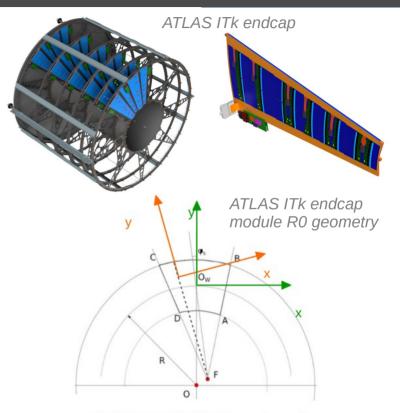


Estimated CPU resources needed for event processing at ATLAS

More complicated detector geometry

- For example, new built-in **radial strips** for the ATLAS Phase II Inner Tracker (ITk) Strip endcap
- And then we got lost in the complexity with various coordinate transformations
 - The <u>ATLAS Software (Athena)</u> release was long broken for ITk before 2018
 - The reconstruction software (<u>Eutelescope</u>) for the ITk module test beam was not usable





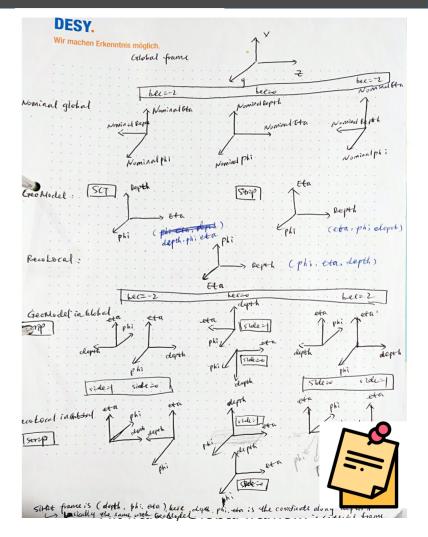
Measurement frame varies with the strip orientation& different to local sensor frame

And we also have the Geant4 Hit frame, Global frame...

More complicated detector geometry

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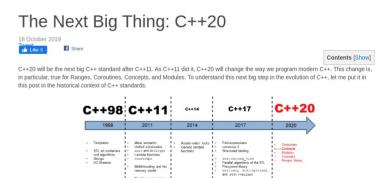
Lots of efforts to put things in order after sorting out the transformations!



Outdated software framework

- Tracking software used at experiments are often developed before first commission data, i.e. ~tens of years old
 - Old design
 - Poor portability
 - Maintenance is a painstaking!
- Optimization of current old software is never easy
- New technology and architecture is there!



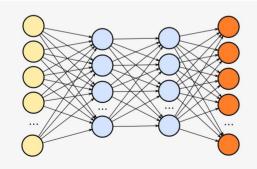


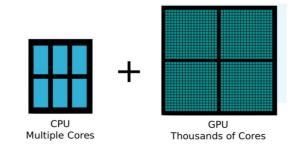
How to achieve accurate, efficient and fast tracking for various detectors?

I. Develop a new high performant common tracking toolkit II. Parallelization, acceleration and ML

The A Common Tracking Software project

- To prepare a modern open-source experimentindependent tracking toolkit for current and future detectors based on LHC tracking experience
 - Targeting at HL-LHC, but also for Belle-II, FASER, sPHENIX, ALICIE, EIC, CEPC...
- To provide an open-source R&D platform to explore new techniques, parallelization and acceleration

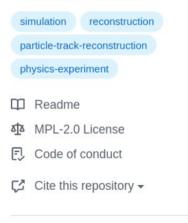






Experiment-independent toolkit for (charged) particle track reconstruction in (high energy) physics experiments implemented in modern C++

About



Latest

Releases 89

v15.0.0

19 days ago

The ACTS developers team

- 10~15 active developers on Core project
 - ATLAS heavy, but increasing external contribution





ACTS fosters collaboration

- World-wide users from particle and nuclear physics, collider and non-collider experiments
 - >10 experiments
 - ATLAS, Belle-II, ALICE, sPHENIX, FASER, EIC, CEPC, LUXE, PANDA, Muon Collider ...
 - >15 institutes
 - CERN, LBNL, ORNL, UC Berkeley, Stanford University, DESY, Universite at Bonn...
 - ~45 forks of the acts repository
- Regular/irregular discussion between developers and experiment users
 - ATLAS, FASER, sPHENIX, ALICE, EIC...



ACTS design

- Modern C++ 17 concepts
- Highly-templated design
 - Detector and magnetic field agnostic
- Strict thread-safety to facilitate concurrency
- Supports for contextual condition data
 - Concurrent event execution with different Geometry/Calibration/Magnetic field in flight
- Minimal dependency (Eigen)
- Highly configurable for usability
- And well-documented !

https://acts.readthedocs.io/en/latest/



Docs » Acts Common Tracking Software

C Edit on GitHub

Acts Common Tracking Software

Acts is an experiment-independent toolkit for (charged) particle track reconstruction in (high energy) physics experiments implemented in modern C++.

The Acts project provides high-level track reconstruction modules that can be used for any tracking detector. The tracking detector geometry description is optimized for efficient navigation and fast extrapolation of tracks. Converters for several common geometry description packages are available. In addition to the algorithmic code, this project also provides an event data model for the description of track parameters and measurements.

Key features:

- A tracking geometry description which can be constructed manually or from TGeo and DD4Hep input.
- · Simple event data model
- Implementations of common algorithms for track propagation and fitting.

How-to guides

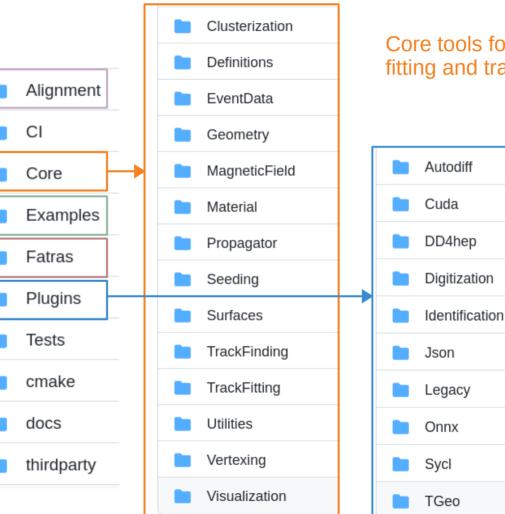
- Run the FAst TRAck Simulation
- ACTS Material Mapping Tutorial
- Run the seeding example
- Run the SYCL seed finding tests
- Run the truth tracking examples
- Run the CombinatorialKalmanFilter (CKF) tracking example
- ACTS Vertexing Tutorial Example: Adaptive Multi-Vertex Finder (AMVF) Pythia8
- Run the alignment examples

The tracking tools in ACTS

An alignment prototype

A light-weight test framework with application examples

A fast simulation engine



Core tools for track propagation, track fitting and track finding, vertexing...

Plugins to support R&D on new techniques!

The tracking tools in ACTS

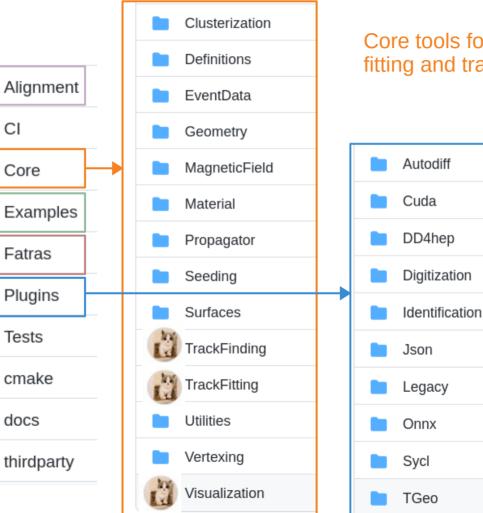
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An alignment prototype

A light-weight test framework with application examples

A fast simulation engine



Core tools for track propagation, track fitting and track finding, vertexing...

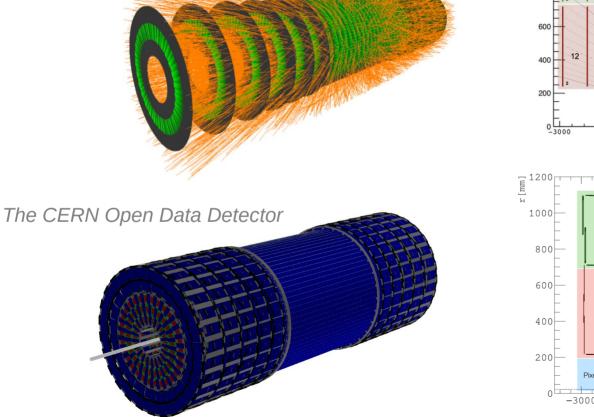
Plugins to support R&D on new techniques!

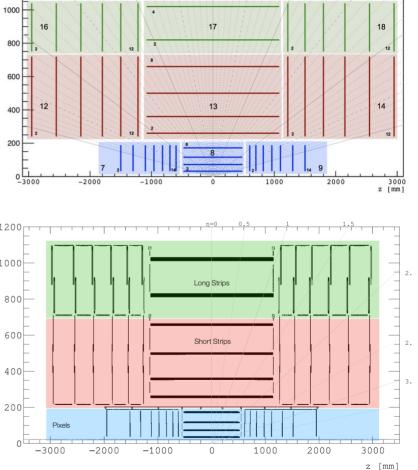
I'll be happy to help there!

The detectors used for development

r [mm]

The TrackML detector



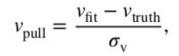


ACTS tracking performance: track parameters estimation

6000 Data Data 5000 Data 5000 Gaussian fit: $\mu = 0.03$. $\sigma = 1.04$ Gaussian fit: $\mu = -0.00$, $\sigma = 0.94$ Gaussian fit: $\mu = -0.03$. $\sigma = 1.05$ 5000 4000 4000 4000 Events/ 0.1 0005 0.1 0.1 0008 0.1 Events/ (000£ Events/ 000 2000 1000 1000 1000 0 -2 2 -2 -2 Ó 2 -4 0 -4 0 2 -4 Δ $pull(d_0)$ $pull(z_0)$ $pull(\phi)$ Data Data Data 5000 5000 5000 Gaussian fit: $\mu = -0.00$, $\sigma = 1.01$ Gaussian fit: $\mu = -0.04$, $\sigma = 1.00$ Gaussian fit: $\mu = -0.00$, $\sigma = 1.00$ 4000 4000 4000 Events/ 0.1 0000 0000 Events/ 0.1 0000 0002 Events/ 0.1 0000 0002 1000 1000 1000 0 0 -2 -2 -2 -4 0 2 -4 0 -4 0 2 $pull(\theta)$ pull(q/p)pull(t)

Pull of track parameters represented at the perigee

<u>arXiv:2106.13593v1</u>



Single muon 400 MeV < pT < 100 GeV, |eta|<2.5 TrackML detector, Bz= 2T

Improve tracking precision with second-order KF

- The (extended) KF used in HEP is optimal for linear system
- Tracking precision is degraded by significant non-linear effects with large incident angle

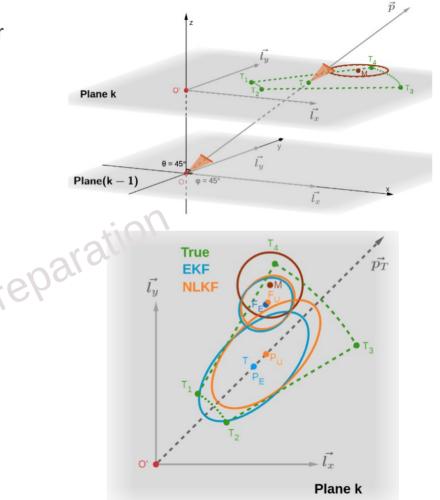
EKF: mean= 0.14, RMS=1.31

NLKF: mean= -0.13, RMS=0.94

Events/ 0.2

1000

500

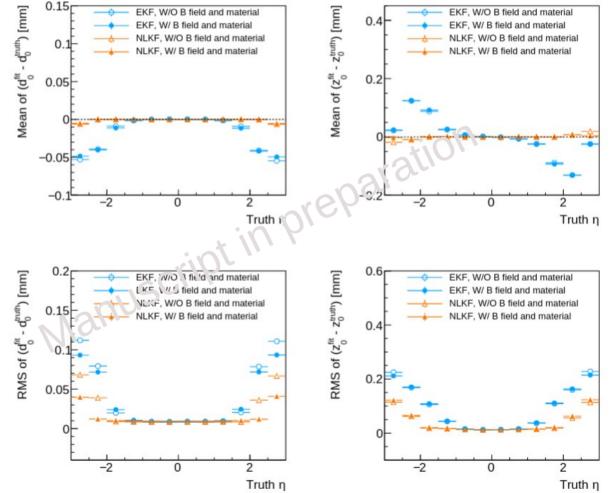


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Improve tracking precision with second-order KF

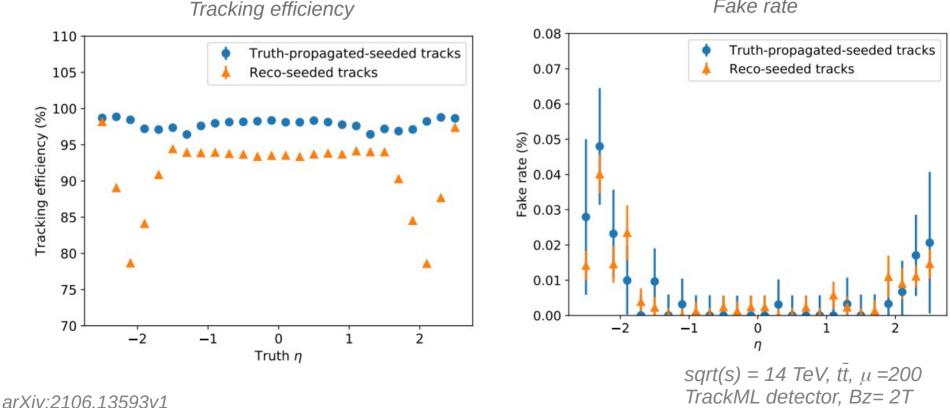
- Application of second-order correction for KF in HEP for the first time!
 - Corrects the bias and improves resolution of track parameters significantly!
- The implementation will be ready for deployment in ACTS soon!

Single muon (20<pT<100 GeV) Open Data Detector, solenoidal Bz= 2T (ATLAS-like)



ACTS tracking performance: efficiency and fake rate

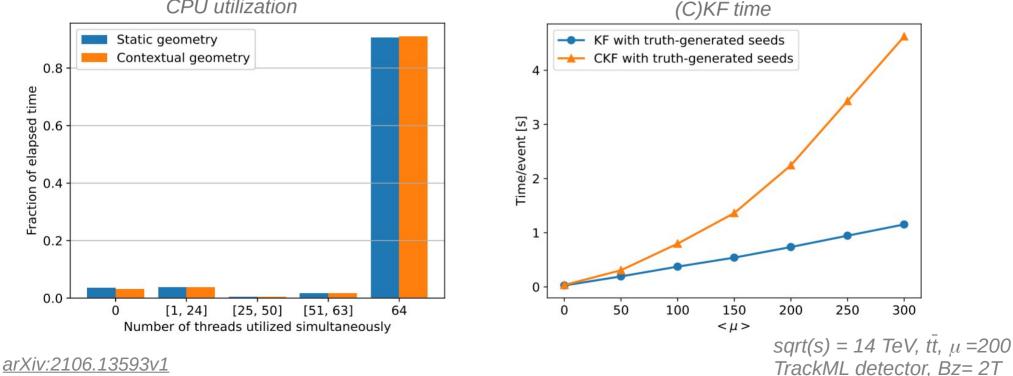
>99% tracking efficiency and <0.01% fake rate for tt at $<\mu>=200$ (~3000 charged tracks/event) •



Fake rate

ACTS CPU performance

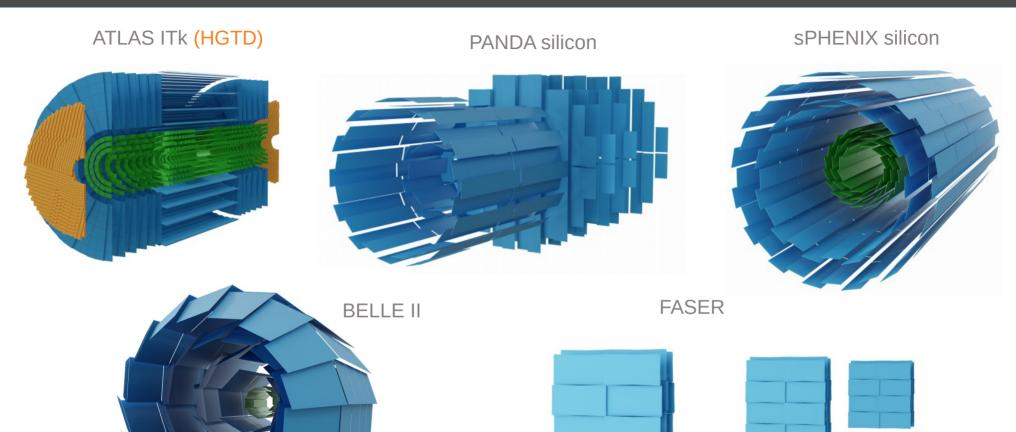
- Efficient CPU utilization even with contextual geometry •
- Pure track fitting time ~ 0.2 ms/ track for ~15 detector layers with a single thread •



CPU utilization

ACTS application examples

arXiv:2106.13593v1



ACTS application example: Muon Collider

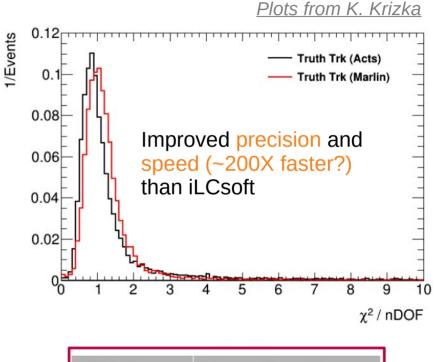


Great potential for discovery in the multi-TeV energy range!

EW measurement, Higgs couplings, new resonances DM search ...

Muon Collider Detector

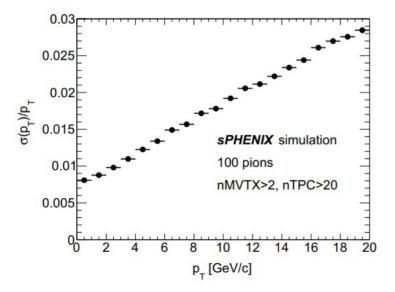
Hit density is 10x HL-LHC due to Beam Induced Background (BIB)

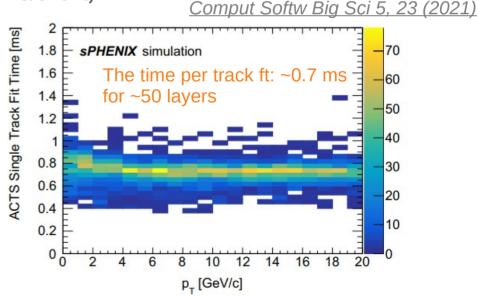


Fit Library	Execution Time		
ACTS	0.5 ms / evt		
iLCsoft	100 ms / evt		

ACTS application example: sPHENIX SPHENIX

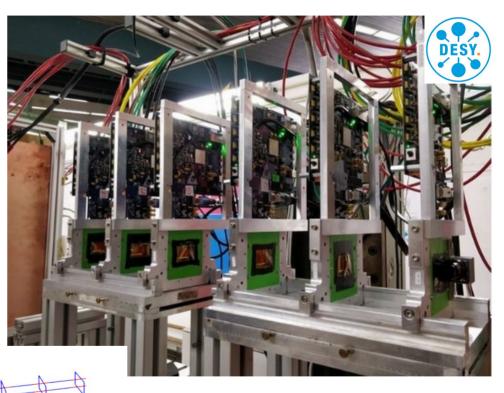
- Study quark-gluon-plasma, partonic structure of protons and nuclei in p+p, p+Au, Au+Au
 - Au+Au collision produces ~1000 tracks/ event
- ACTS provides good tracking resolution needed to resolve high momentum jets
 - $\Delta p/p \leq 0.2\%$ ·p (GeV) for pT > 10 GeV tracks
- ACTS provides X8 faster tracking than GenFit package
 - Total tracking time is 10 s/event (fitting time: ~1 s/event)





ACTS application example: ADENIUM beam telescope

- Beam telescope is a key instrumental tool for particle detector prototyping
- Combined tracking fitting and finding with CKF much ease the tracking process
- Good time performance allows online track reconstruction and visualization
 - Event processing rate up to 20 kHz in a single thread!



ADENIUM beam telescope developed by Y. Liu

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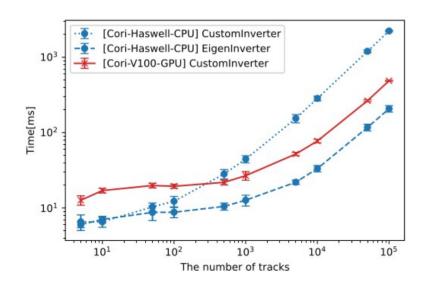
Online visualization of two tracks in one trigger

How to achieve accurate, efficient and fast tracking for various detectors?

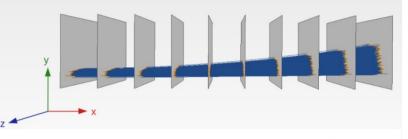
I. Develop a new high performant common tracking toolkit II. Parallelization, acceleration and ML

The GPU-accelerated track fitting

- ALICE, LHCb. CMS are exploring GPU for HLT trigger
- The first look at heterogeneous computing in ACTS was porting a full KF to GPU
 - ~ 4.5X speed gain for >1000 tracks!
 - But not detector agnostic yet



<u>Comput Softw Big Sci 5, 20</u> <u>See Poster at ACAT21</u>



Sys.	CPU	S×C×T	Clock rate (GHz)	Mem. (GB)
1	Intel Xeon E5-2698 v3 (Cori-Haswell-CPU)	2×16×2	2.30	128
1	Intel Xeon Phi 7250 (Cori-KNL-CPU)	1×68×4	1.40	96
2	Intel Xeon Gold 5115 (NAF-SL-CPU)	2×10×2	2.40	376

Sys.	GPU	FP32 cores	FP64 cores	Clock rate (GHz)	Mem. (GB)
1	GV100-SXM2 (Cori-V100-GPU)	5120	2560	1.53	16
2	GP100-PCIe (NAF-P100-GPU)	3584	1792	1.48	16
2	GV100-SXM2 (NAF-V100-GPU)	5120	2560	1.53	32

The current ACTS parallelization R&D landscape

- Active on-going development towards a full track chain on GPU in ACTS community
 - Needs general solutions for difficulties with GPU: C++ STL containers and algorithms not usable, polymorphism not supported ...

```
      Vectorised data model base and helper classes.

      ● C++
      ☆ 8
      ☆ MPL-2.0
      ♀ 7
      ⊙ 2 (1 issue needs help)

      detray
      Public

      Test library for detector surface intersection

      ● C++
      ☆ 4
      ☆ MPL-2.0
      ♀ 4
      ⊙ 5 (3 issues need help)
```

Designing **STL-like containers** for both CPU and GPU

Geometry navigation **without runtime polymorphism** based on indiced surfaces (boundaries, transformations, material...)

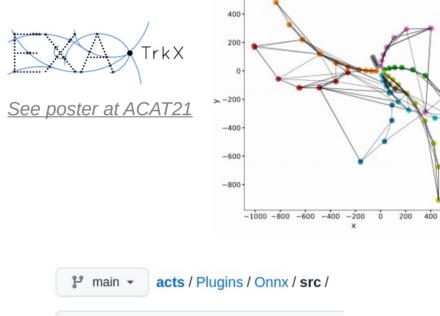
Machine Learning-based tracking algorithms

- ML is widely deployed in tracking domain
 - GNN for track finding (e.g. Exa.TrkX)
 - Evolutionary algorithm for parameters tuning
 - DNN based track classification
 - KNN for surface prediction

. . .

 The microsoft <u>ONNX</u> plugin was implemented in ACTS to allow deployment of ML solutions





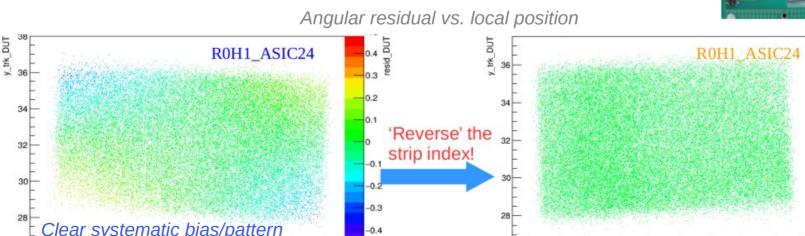


Track-based detector alignment

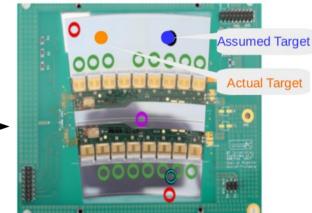
Tracking can help with detector alignment

- Alignment is necessary to provide accurate description of detector placement (translation + rotation)
- Misalignment will deteriorate the track resolution, but tracking can notice and correct the misalignment
 - e.g. a mistake of moving DUT to wrong target, detected by the dedicated tracking for radial strips

x trk DUT



First ATLAS ITk Endcap strip R0 module at testheam in 2017



0.4

0.3

-0.1

-0.2

-0.4

trk DUT

How to do track-based alignment?

• Estimation of alignment parameter α that is common for a sample of tracks by minimizing the total χ^2 w.r.t. to track parameters x and α :

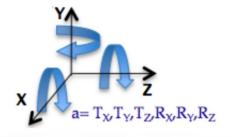
$$\chi^2 = \sum_i \chi_i^2 = \sum_i [\vec{m_i} - \vec{h_i}(\vec{x_i}(\vec{\alpha}), \vec{\alpha})]^T V^{-1}[\vec{m_i} - \vec{h_i}(\vec{x_i}(\vec{\alpha}), \vec{\alpha})]$$

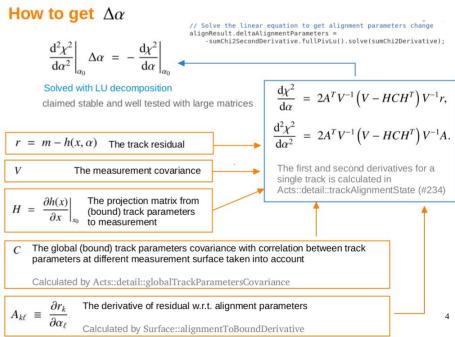
• Found the best α by minimizing the χ^2 , i.e.

$$- \qquad 0 \equiv \frac{d\chi^2}{d\vec{\alpha}} \text{ with } \forall_i \frac{\partial \chi_i^2}{\partial \vec{x_i}} = 0$$

- Needs to obtain $\Delta \alpha$ via solving:

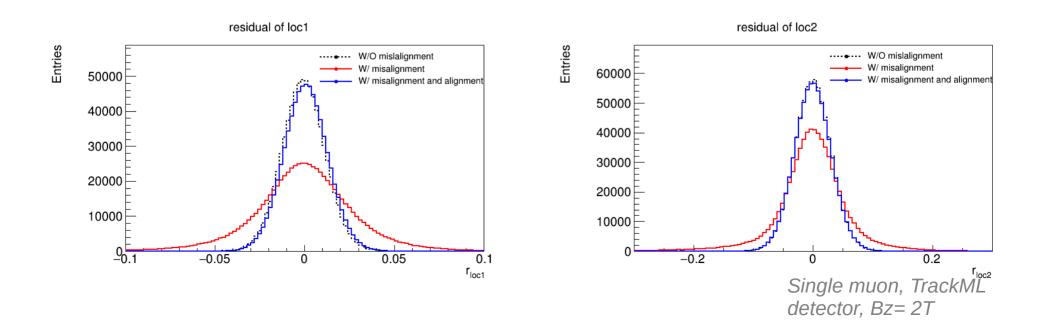
$$\frac{d^2\chi^2}{d^2\vec{\alpha}}\mid_{\vec{\alpha_0}}\Delta\vec{\alpha} = -\frac{d\chi^2}{d\vec{\alpha}}\mid_{\vec{\alpha_0}}$$





Using KF for track-based detector alignment

- KF-based tracking is commonly used, e.g. ATLAS and CMS experiments
 - KF-based alignment is more straightforward than χ^2 fitter based alignment
- A KF-based alignment prototype in ACTS has been developed and looks promising!



Summary

- Tracking is pivotal to reconstruction in HEP
- Tracking has direct and indirect impact to physics precision
- Tracking is non-trivial and can be more challenging at future HEP experiments
- Modern performant common tracking software (ACTS) is being developed and gets worldwide users from >10 experiments

- Tracking is pivotal to reconstruction in HEP
- Tracking has direct and indirect impact to physics precision
- Tracking is non-trivial and can be more challenging at future HEP experiments
- Modern performant common tracking software (ACTS) is being developed and gets worldwide users from >10 experiments
- Outlook
 - There are still tools to develop and optimize in ACTS
 - Interplay with experiment frameworks is the key to make a real generic toolkit

Collaboration is always welcome!







acts-users@cern.ch

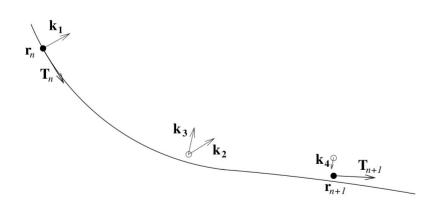






Track parameter propagation

- Magnetic field is usually non-homogeneous.
 - The particle motion equation needs to be solved numerically
- The fourth-order Runge-Kutta-Nyström method for



E. Lund et al. 2009 JINST 4 P04001

$$\frac{d^2y(x)}{dx^2} = f(x, y, y')$$

$$k_{1} = f(x_{n}, y_{n}, y'_{n})$$

$$k_{2} = f\left(x_{n} + \frac{h}{2}, y_{n} + \frac{h}{2}y'_{n} + \frac{h^{2}}{8}k_{1}, y'_{n} + \frac{h}{2}k_{1}\right)$$

$$k_{3} = f\left(x_{n} + \frac{h}{2}, y_{n} + \frac{h}{2}y'_{n} + \frac{h^{2}}{8}k_{1}, y'_{n} + \frac{h}{2}k_{2}\right)$$

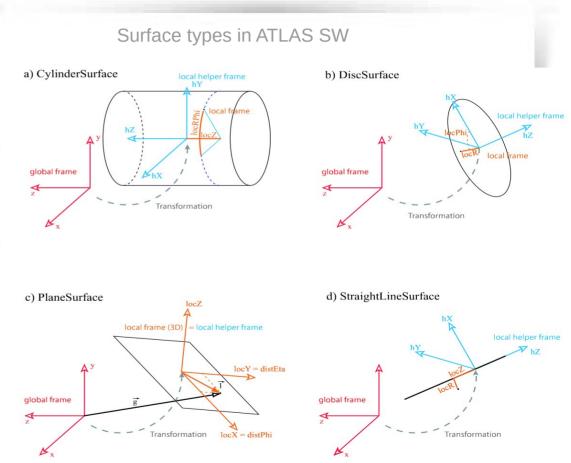
$$k_{4} = f\left(x_{n} + h, y_{n} + hy'_{n} + \frac{h^{2}}{2}k_{3}, y'_{n} + hk_{3}\right)$$

$$y'_{n+1} = y'_{n} + \frac{h}{6}(k_{1} + 2k_{2} + 2k_{3} + k_{4})$$

$$y_{n+1} = y_{n} + hy'_{n} + \frac{h^{2}}{6}(k_{1} + k_{2} + k_{3})$$

ACTS Surface

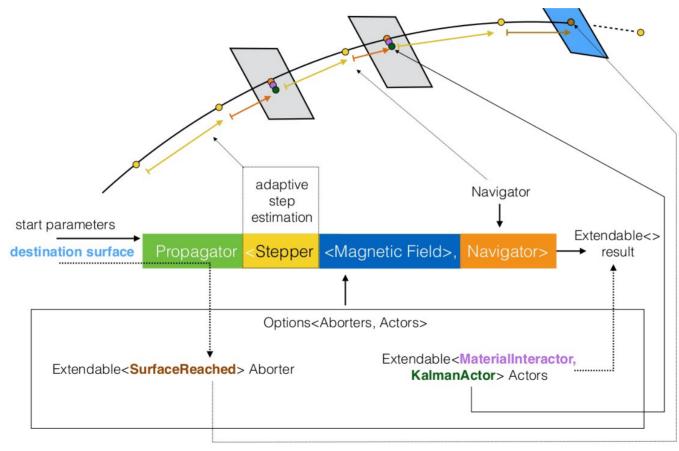
- Acts::Surface is the key component of tracking geometry
 - Surface concepts are largely transcribed from ATLAS SW
- Different concrete surfaces have different local coordinate definitions and shapes
 - Shape is described by Acts::SurfaceBounds



ACTS Track parameter propagator interface

Integrating particle transport & geometry navigation

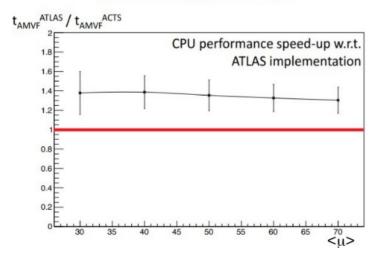
Highly-templated design emphasizing on speed and customizability

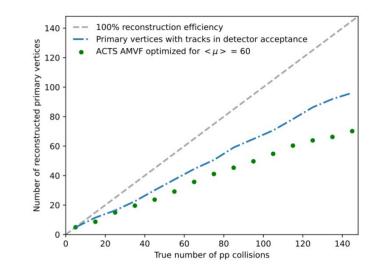


ACTS vertexing

- Iterative fitting-after-finding
 - Iterative Vertex Finder (IVF) (used at ATLAS Run-2)
- Finding-through-fitting
 - Adaptive Multi-Vertex Finder (AMVF) (to be used at ATLAS Run-3)

AMVF timing performance





See B. Schlag's slides