

Track reconstruction with ACTS in modern high-energy physics experiments

Gundolf Haase, Alexander J. Pflieger, Andreas Salzburger



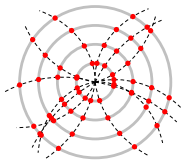
22nd of September 2026

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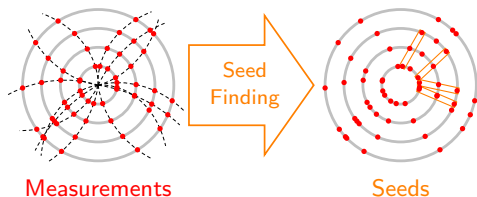
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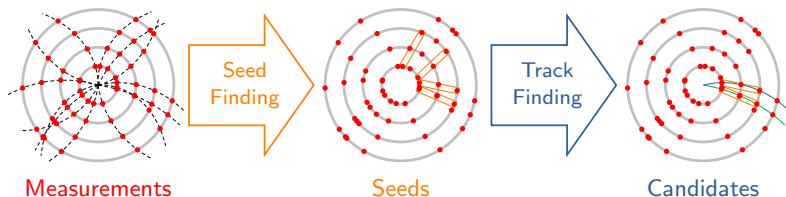


Measurements

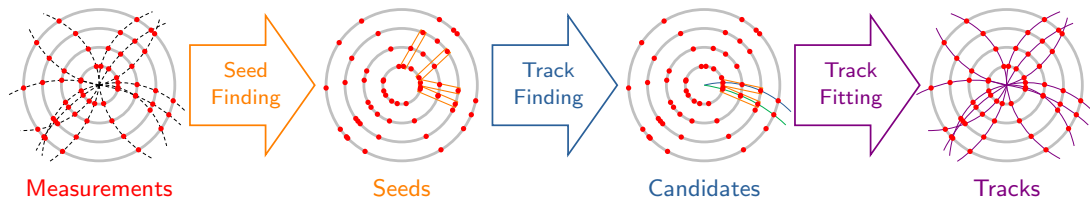
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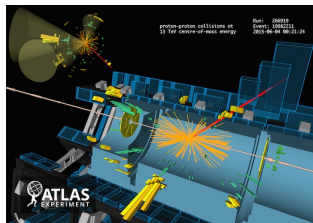


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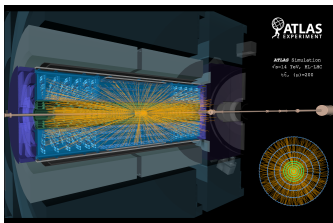
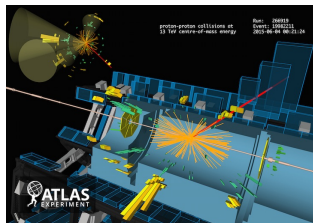
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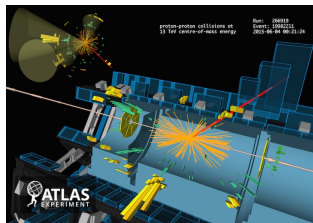
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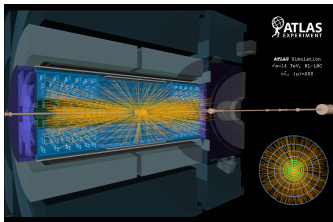


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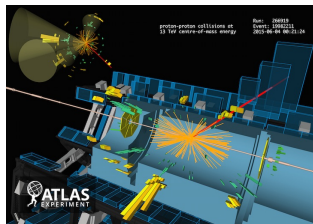


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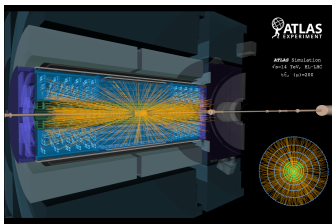
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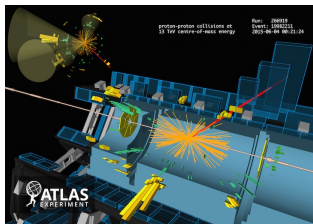


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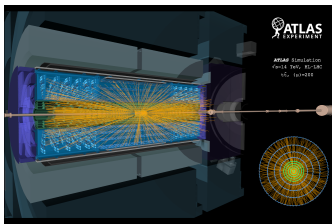
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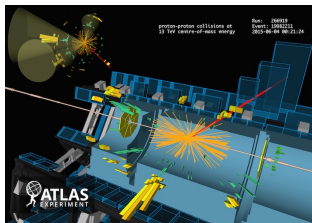


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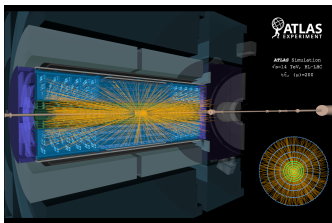
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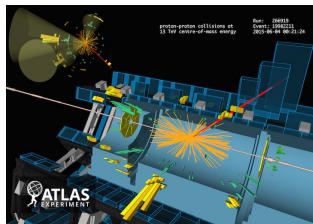


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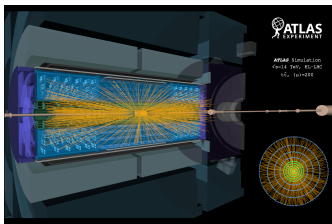
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- ▶ Natively single-threaded

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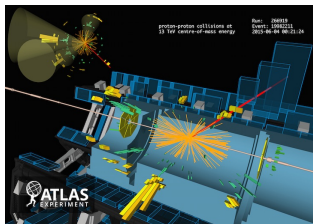


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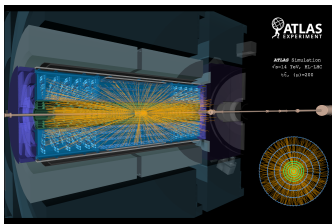
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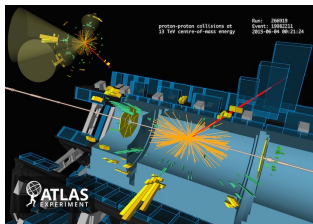


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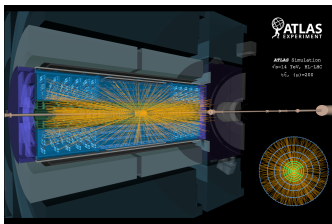
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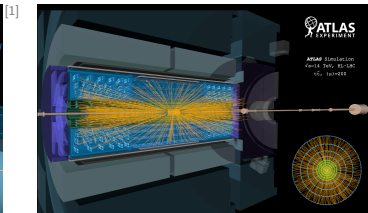
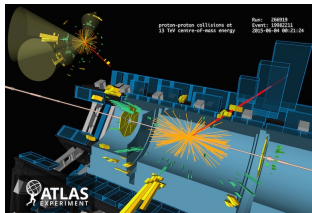
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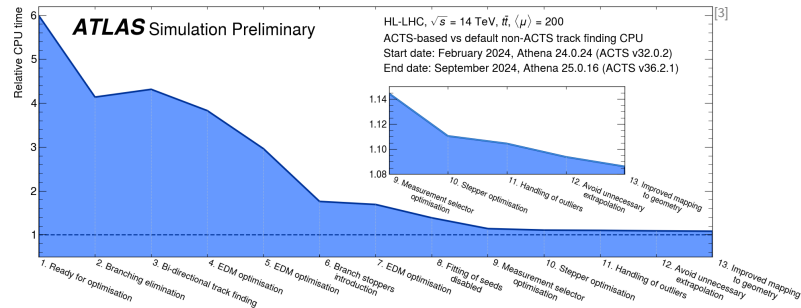
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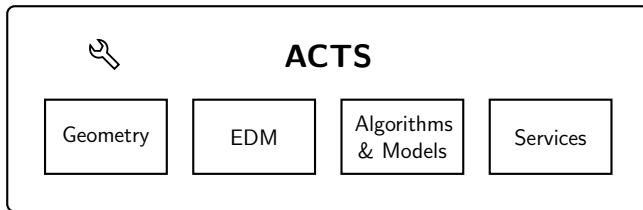
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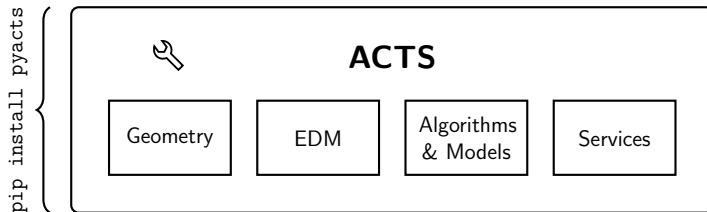


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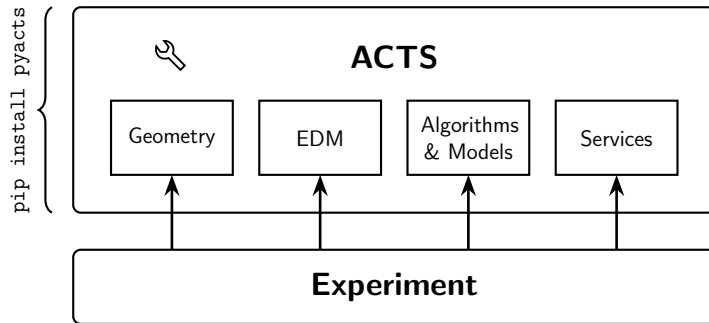
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 [3] ATL-PHYS-PUB-2024-017

ACTS

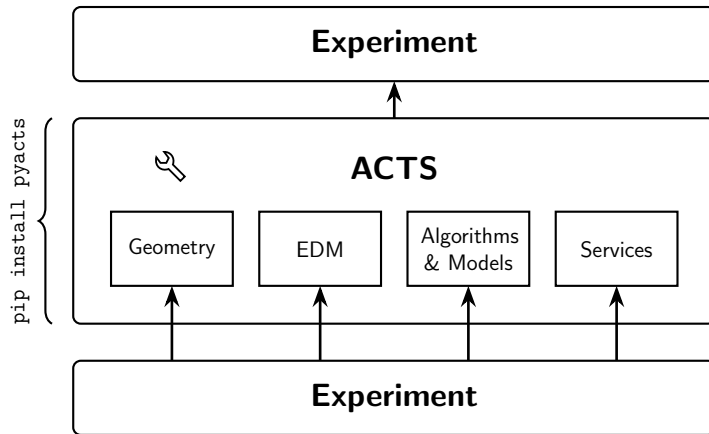




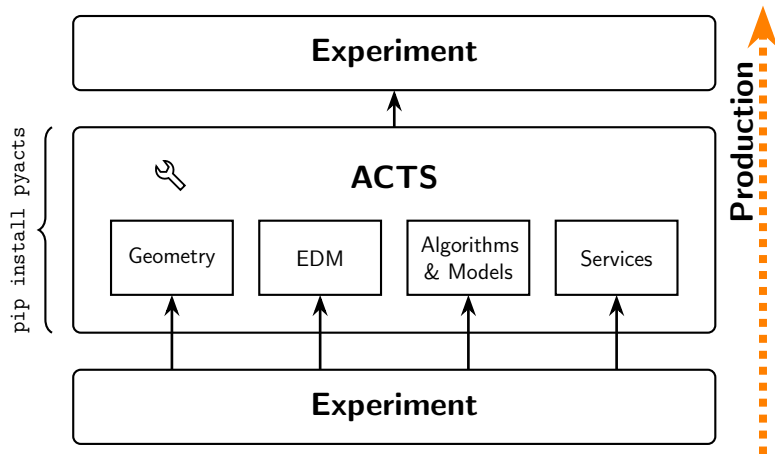
<https://github.com/acts-project/acts/>



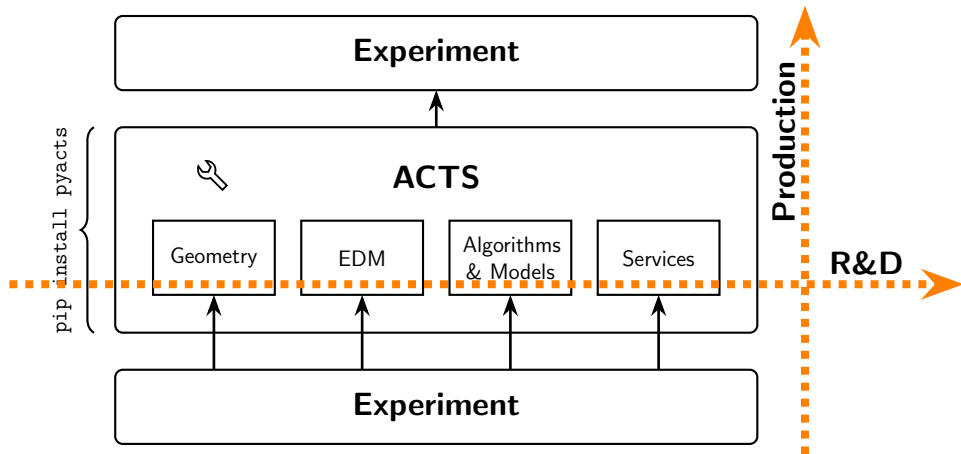
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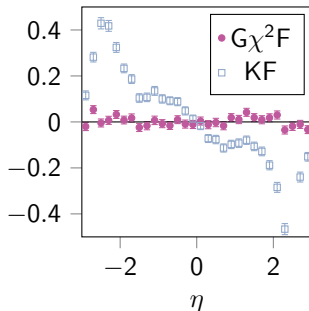


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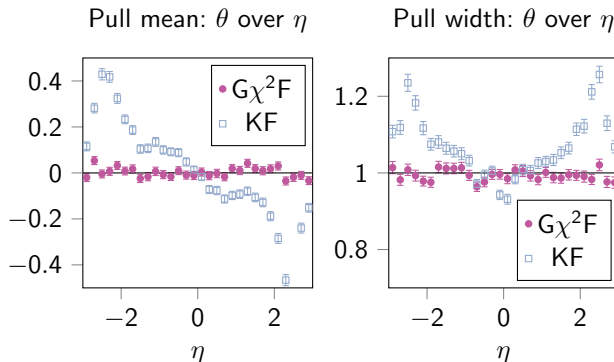
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- ▶ $pT \in [1, 100]$ GeV/c, randomised charge (to reduce biases)

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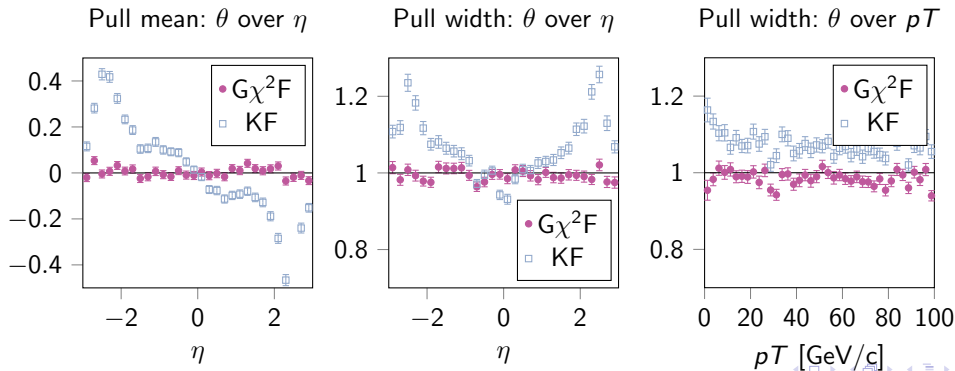
Pull mean: θ over η

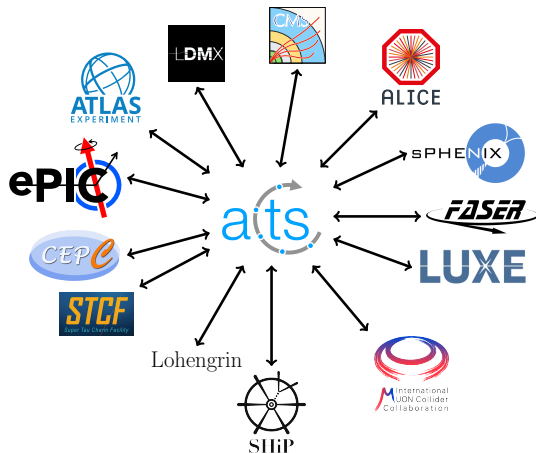


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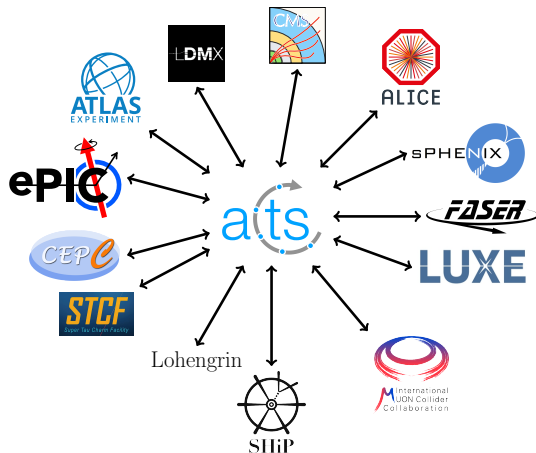


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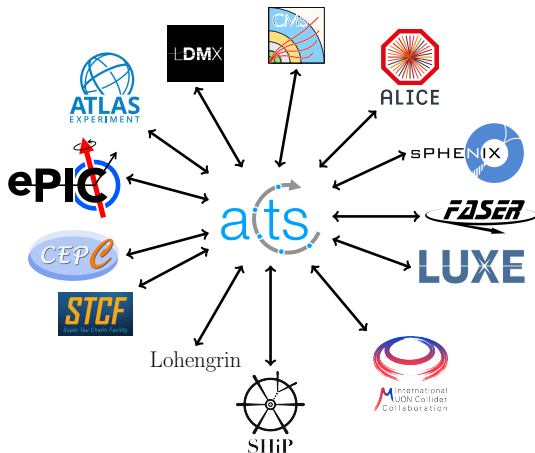




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- ▶ Different detectors expose hidden assumptions
- ▶ Share development and maintenance

- ▶ Can machine learning reduce candidate combinations?

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Towards a realistic track reconstruction algorithm based on graph neural networks for the HL-LHC

Implementation of ACTS for STCF track reconstruction

Auto-tuning capabilities of the ACTS track

Track reconstruction as a service for collider physics

Haoran Zhao^a Yuan-Tang Chou^a Yao Yao^b Xiangyang Ju^c Yongbin Feng^d William Patrick McCormack^e Miles Cochran-Branson^f Jan-Frederik Schulte^b Miaoyuan Liu^b Javier Duarte^f Philip Harris^e Shih-Chieh Hsu^g Kevin Pedro^g Nhan Tran^g

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ABSTRACT: Optimizing charged-particle track reconstruction algorithms is crucial for efficient event reconstruction in Large Hadron Collider (LHC) experiments due to their significant computational demands. Existing track reconstruction algorithms have been adapted to run on massively parallel coprocessors, such as graphics processing units (GPUs), to reduce processing time. Nevertheless, challenges remain in fully harnessing the computational capacity of coprocessors in a scalable and non-disruptive manner. This paper proposes an inference-as-a-service approach for particle tracking in high energy physics experiments. To evaluate the efficacy of this approach, two distinct tracking algorithms are compared: *PythiaTrack*, a rule-based algorithm, and *PythiaTrack*, a machine learning-based

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traccc: GPU track reconstruction library for HEP experiments

Paul Gessinger¹, Heather M. Gray^{2,3}, Attila Krasznahorkay⁴, Charles Leggett³, Joana Niemann¹, Andreas Salzburger¹, Stephen Nicholas Swatman¹, and Beomki Yeo^{2,3,*}

¹European Organization for Nuclear Research, Meyrin, 1211, Switzerland

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³Physics Division, Lawrence Berkeley National Laboratory, Berkeley, CA 94720, USA

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Status of ACTS Tracking for FCC detectors

- ▶ ACTS is ATLAS run-4 standard and deployed

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- ▶ R&D: machine learning, GPU tracking, future detectors, ...

Extra Slides



Enhancing CMS Passive Material Modeling in Track Reconstruction Using ACTS

Lorenzo Damenti

ACTS Workshop, 20-24 April 2026, NIKHEF

Proof of Concept

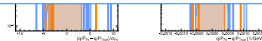
- First test: Perform track refits on simulated muons ($p_T = 100\text{ GeV}$)
- Purpose: Validate geometry built and identify major showstoppers
- Note:** Test performed using all algorithms from ACTS (i.e. propagation and refit)
- Results show a good agreement between ACTS and CMSSW



Problem: Potential improvements in passive material treatment could be hidden by non-optimized ACTS algorithms for CMS
Solution: Import only the ACTS passive material treatment into a new CMSSW propagator

Advantages:

- Facilitates extending ACTS to other reconstruction stages
- Simplifies ACTS adoption within the CMS community



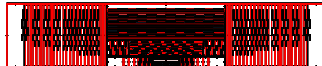
- Next Step:** decrease the p_T to probe passive material effects? Yes, but...

CMS Tracking Geometry in ACTS

- All sensitive surfaces from a TrackerGeometry in CMSSW are selected and converted into ACTS DetectorElements (DetEl)s
- Note:** $detID_{CMSSW} \leftrightarrow detID_{ACTS}$
- DetEl's are organized in tracking layers; **Note:** not always tracking layers correspond to physical layers



- Layers are arranged into volumes following a hierarchical structure
- Full CMS blueprint:



- Open Points:**

- Is the number of material layers too high?
- Is the layers placement correct?
- Is the binning appropriate?

Geometry Optimization

21/04/26

ACTS Workshop 2026

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Summary and Next Steps

- Description of passive material effects in CMSSW is suboptimal $\rightarrow$ verify if it can be improved by importing the passive material treatment approach adopted by ACTS defining a new ACTS propagator within the CMS Data Analysis Framework
- Build (and validate) the CMS Geometry within ACTS ✓
- Validate material maps used to decorate material layers ✓
- Write a new CMSSW propagator using ACTS ✓
- Validate the new propagator:
 - Compare refit performance (pulls and residuals distributions) *Ongoing*
 - Compare Track Efficiency (Using ACTS propagator in all track reco steps)
- Reconstruction of $K_S^0 \rightarrow \pi^+ \pi^-$ decays and comparison between ACTS and CMSSW



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- ▶ Full algorithm
- ▶ Already explored by experiments
- ▶ Austrian Doctoral Student Programme
- ▶ PhD at Uni Graz at CERN

Symmetries & Unambiguous Ambiguities

Track Fitting with the Global χ^2 Fitter in ACTS

Alexander J. Pfeiffer¹, Paul Gessinger-Befurt¹, Gundolf Haase²
¹ Uni Graz, Austria | CERN

Open Data Detector

Let's compare the performance of the $G_{\chi^2}F$ and the **Kalman Filter Fitter (KF)** on the ODD.
 We start with $10^5 \mu^-$.
 To make it more realistic vary the energy between $1 - 100 \text{ GeV}$ and include **material effects**.

Results:

- Drop unreachable measurements
- Unstable tracks improve
- Fit quality independent of direction and momentum
- Slower: Propagation is done multiple times

(The ODD is symmetric)

The $G_{\chi^2}F$ can just ignore faulty measurements.

Drift Tubes

DTs provide very precise spatial information, but each drift measurement has an **inherent left/right ambiguity**.

We only know how far the particle passed from the wire, not on which side.

The **z-position** (along the wire) is determined by additional strip detectors mounted above and below the tubes.

But how does the $G_{\chi^2}F$ know which side to choose?

- Prediction: Ignore which side we are passing
- Calculate residuals with absolute values
- Additional strip / pixel detectors restrict the possible trajectory
- The $G_{\chi^2}F$ naturally selects the correct configuration, and the ambiguity disappears

(Drift tube measurements are ambiguous)

alexander-julian.pfeiffer@cern.ch
github.com/AJPfeiffer/acts

Connecting the Dots
 10-14 Nov 2025, Tokyo

<https://indico.cern.ch/event/1499357/contributions/6628643/>