ATLAS 4D Tracking for HL-LHC VII Reunião Geral - Projeto Temático

Rodrigo Estevam de Paula

Marco Aurelio Lisboa Leite

Vitor Heloiz Nascimento

15 October, 2024



ATLAS preparations for HL operation: HGTD

• A High Granularity Timing Detector (HGTD) will be installed in the forward region of the ATLAS detector to provide track timing information that will complement the reconstruction





Simulation of Z \rightarrow µµ with < µ > = 140, as seen by the ATLAS experiment

New reconstruction methods that are adapted to the new detector and meet performance requirements from the HL-LHC are necessary!

Simulating ATLAS with ACTS

- Using ACTS to simulate the track reconstruction with ITk and HGTD
 - More details on my previous presentation
- Managed to include ITk+ HGTD geometry
- Our first objective was to evaluate the performance of CKF after including HGTD time measurements



Including HGTD time smearing at geometry files

- At the digitization step, ACTS uses a geometry config file to simulate smearing of measurements
- We included the time parameter at volumes 2 and 25, which represent HGTD
 - \circ Other volumes just measure I₀ and I₁

 $\vec{x} = (l_0, l_1, \phi, \theta, q/p, t)^T$

•••

```
"acts-geometry-hierarchy-map": {
    "format-version": 0,
    "value-identifier": "digitization-configuration"
"entries": [
        "volume": 2,
        "value": {
            "smearing":
                    "index": 0,
                    "mean": 0.0,
                    "stddev": 0.37527767,
                    "type": "Gauss"
                },
                    "index": 1.
                    "mean": 0.0,
                    "stddev": 0.37527767.
                    "type": "Gauss"
                },
                    "index": 5,
                     "mean": 0.0.
                    "stddev": 0.37527767,
                    "type": "Gauss"
```

CKF performance evaluation on Particle guns

- Incorporated HGTD time measurements on the CKF reconstruction method
- Below are the results for the simple scenario of a particle gun
 - Muons are generated at the coordinate (0,0,0,0) and shot at the detector with a variable angle $\eta \sim U(-4,4)$



Simulating HL-LHC beam with ACTS Pythia8

- Managed to simulate the same scenario as the TDR
 - z0 ~ N(0,50 mm), t0 ~ N(0,175ps)
- Simulated ttbar events with 1000 events







CKF Performance evaluation: Residue plot

- It is possible to isolate tracks tagged by HGTD with the reconstruction covariance matrix
- Tracks without HGTD tagging rely on first estimate for time reconstruct, which has a high error associated with.
- When a track reaches HGTD, the error drops significantly because of the low error attributed to the measure
 - In the plot aside, the covariance threshold to separate the tracks (err_eT) was set to 1000



AMVF Vertex time reconstruction

- To understand the impact of CKF on the global t₀ reconstruction we added a step to estimate primary vertex positions using AMVF
 - Detailed explanation of AMVF on backup
- Estimates with high residual can mostly be filtered by the time error parameter on the covariance matrix (internal parameter of AMVF)
 - Work point choice of purity x sensibility
- After filtering out high error estimates, a Gaussian fit was made to get t_{0rec} resolution



Simulation of sensor deterioration

- We simulated the sensor deterioration with increasing integrated luminosity
- ACTS could be used as a quick test bench before a full implementation in ATHENA
 - Way faster to implement and to test the full reconstruction chain with the changes
- We can use this to evaluate the impact of degradation on the track reconstruction



Presentation at the ACTS Developers Workshop

- Will present these results at the <u>ACTS Developers Workshop</u>
- The presentation will focus on the performance of CKF time reconstruction using a timing layer detector and the impact of sensor degradation on that vertex reconstruction





ATLAS Qualification Project

- Started an Authorship Qualification Project together with HGTD Simulation and Performance team
- The first goal is to include the simulation of HGTD front end electronics response in the ATHENA framework
 - Simulation of digitization of Time to Digital Converters (TDCs) and calculation of Time of Arival (TOA)
 - \circ \quad Allows for more accurate Monte Carlo simulation of the detector
- Implementation is done, just need to discuss some code organization and changes to downstream components (cluster building)



Next steps

- Now that we have a good understanding of CKF and its performance, we can start the research of new reconstruction methods
- For now, ATLAS has been supporting the GNN tracking method for its, but the chose has yet to be made



We're on the right track 😉



- Extending Tra TrackML ference timing and scaling studies to
- Investigating training and inference performance on lower p_T tracks (i.e. < 1 GeV) and high p_T tracks (i.e. > 10 GeV)
- Investigating performance on large radius tracks and dense track environments
- Direct comparison with combinatorial Kalman filter (current algorithm) efficiency and track parameter resolution



HighRR Lecture Week - Heidelberg University - September 13, 2023

87

ATLAS

Taken from Daniel Murnane lecture [3]

Bonus: Finally finished interal note 😀

- The note can be found in CDS: <u>https://cds.cern.ch/record/2863410</u>
- Documents (some of) our contributions to the commissionament of Phase-I LAr upgrade



Thank you for your attention!

Questions?

Referencias

[1] ATLAS Collaboration. Observation of a new particle in the search for the Standard Model Higgs boson with the ATLAS detector at the LHC. Physics Letters B,2021. <u>https://doi.org/10.1016/j.physletb.2012.08.020</u>.

[2] M. Gullstrand, and S.Maraš. "Using Graph Neural Networks for Track Classification and Time Determination of Primary Vertices in the ATLAS Experiment" (Dissertation). Disponível em <u>http://urn.kb.se/resolve?urn=urn:nbn:se:kth:diva-288505</u>

[3] Daniel Murnane. Full-length tutorial on Tracking with Graph Neural Networks (Sep 2023, Heidelberg). Available at: <u>https://exatrkx.github.io/</u>

[4] ATLAS Collaboration, "Technical Design Report for the ATLAS ITk Pixel Detector", Tech. Rep. ATL-COM-ITk-2018-019, CERN, Geneva, 2018.Disponível em: <u>https://cds.cern.ch/record/2310230</u>

[5] ATLAS Collaboration, "Technical Design Report: A High-Granularity Timing Detector for the ATLAS Phase-II Upgrade". Technical report, CERN, Geneva, 2020. Disponível em: <u>https://cds.cern.ch/record/2719855/files/ATLAS-TDR-031.pdf</u>

[6] Paul Gessinger-Befurt. *Development and improvement of track reconstruction software and search for disappearing tracks with the ATLAS experiment*, 2021. Presented 30 Apr 2021.

[7] ATLAS Collaboration. *ACTS documentation*. Disponivel em: <u>https://acts.readthedocs.io/en/latest/index.html</u>



AMVF Time reconstruction

AMVF overview

- I managed to include time information in the AMVF reconstruction
- The following slides will be an overview of the method in order to understand its performance
- The process can be divided into three steps:
 - Vertex seeding: Gaussian Track Density
 - Vertex finding: AMVF
 - Vertex fitting: Kalman Filter updater
- The steps are looped until all (valid) tracks are assigned to a vertex



- Reference for this section:
 - ATLAS Collaboration (2019). Development of ATLAS Primary Vertex Reconstruction for LHC Run 3 [White paper]. CERN.

Vertex seeder

• The seeding step establishes first estimates for primary vertices positions



- Most of the vertex seeders project tracks to origin and evaluate density distribution
 - The peaks of the distribution will be the seeds for vertices

Gaussian Track Density

• The density of track origin can be represented by a multi-variate Gaussian distribution:

$$P(r,z) = \frac{1}{2\pi\sqrt{|\Sigma|}} e^{-\frac{1}{2}((r-d_0),(z-z_0))^T \Sigma^{-1}((r-d_0),(z-z_0))}. \qquad \Sigma = \begin{pmatrix} \sigma^2(d_0) & \sigma(d_0,z_0) \\ \sigma(d_0,z_0) & \sigma^2(z_0) \end{pmatrix}$$

• If we project it to the z-axis, we have the density at that axis. Furthermore, If we consider the distribution to be locally Gaussian, it's possible to do a peak search with steps of size:

$$W(z) = \sum_{i \in \text{tracks}} P_i(0, z). \qquad \Delta z = \frac{W(z)W'(z)}{W'^2(z) - W''(z)W(z)}.$$

• The output will be the position of the peak and the width of the distribution around it

$$z_{max} = \max_{z} W(z) \qquad \qquad \sigma(z) = \sqrt{-\frac{W(z_{max})}{W''(z_{max})}}$$

Adaptive Multi Vertex Finding (AMVF)

- Given a collection of reconstructed tracks and estimates of vertexes, establishes a "compatibility" value for each track-vertex
- The algorithm is adaptive in a sense that vertexes compete for the same track (multiple vertex-tracks weights)
- Iterate over association weights until convergence
- Short paper explaining

Fitting procedure

- Fit all vertexes using the assignment probability as track weights
- Recompute the assignment probabilities using the most recent vertex positions

Weight function

• Having n tracks to be fitted to m vertexes The weight of vertex j to track i is:

$$w_{ij} = \frac{\exp(-\chi_{ij}^2/2\,T)}{\exp(-\chi_{\rm cut}^2/2\,T) + \sum_{k=1}^m \exp(-\chi_{ik}^2/2\,T)}$$

T is a temperature parameter X²cut is a cut-off to suppress not assigned tracks X²ij is the chi2 distance between track and vertex

Kalman Filter Updater

- From the collection of tracks assigned to a vertex originated from the previous step, a Kalman Filter is used to fit the vertex position
- The position of the first deposition (measurement) is used to evaluate the vertex position and momentum of the track (vector state)
- The measurement equation would be:



Fig. 8.1 A vertex fit with four tracks. The parameters of the fit are the vertex v and the momentum vectors p_i ; the observations are the estimated track parameters q_i

 $\boldsymbol{q}_i = \boldsymbol{h}_i(\boldsymbol{v}, \boldsymbol{p}_i), \ i = 1 \dots, n.$

Kalman Filter updater equations

- The measurement equation is linearized so the Kalman Filter can be used
- Check on the book for detailed explanation
- The AMVF weights multiply the inverse of the covariance matrix as to imitate a "significance" of the measurement

$$\boldsymbol{v}_{i} = \boldsymbol{C}_{i} \left[\boldsymbol{C}_{i-1}^{-1} \boldsymbol{v}_{i-1} + \boldsymbol{A}_{i}^{\mathsf{T}} \boldsymbol{G}_{i}^{\mathsf{B}} (\boldsymbol{q}_{i} - \boldsymbol{c}_{i}) \right],$$
$$\boldsymbol{p}_{i} = \boldsymbol{W}_{i} \boldsymbol{B}_{i}^{\mathsf{T}} \boldsymbol{G}_{i} (\boldsymbol{q}_{i} - \boldsymbol{c}_{i} - \boldsymbol{A}_{i} \boldsymbol{v}_{i}),$$

$$\operatorname{Var}[\boldsymbol{v}_{i}] = \boldsymbol{C}_{i} = \left(\boldsymbol{C}_{i-1}^{-1} + \boldsymbol{A}_{i}^{\mathsf{T}}\boldsymbol{G}_{i}^{\mathsf{B}}\boldsymbol{A}_{i}\right)^{-1},$$
$$\operatorname{Var}[\boldsymbol{p}_{i}] = \boldsymbol{W}_{i} + \boldsymbol{W}_{i}\boldsymbol{B}_{i}^{\mathsf{T}}\boldsymbol{G}_{i}\boldsymbol{A}_{i}\boldsymbol{C}_{i}\boldsymbol{A}_{i}^{\mathsf{T}}\boldsymbol{G}_{i}\boldsymbol{B}_{i}\boldsymbol{W}_{i},$$
$$\operatorname{Cov}[\boldsymbol{v}_{i}, \boldsymbol{p}_{i}] = -\boldsymbol{C}_{i}\boldsymbol{A}_{i}^{\mathsf{T}}\boldsymbol{G}_{i}\boldsymbol{B}_{i}\boldsymbol{W}_{i}.$$

update equation for both vertex position and particle momentum

AMVF overview (again)

- The process can be divided into three steps:
 - Vertex seeding: Gaussian Track Density
 - Vertex finding: AMVF
 - Vertex fitting: Kalman Filter updater
- The steps are looped until all (valid) tracks are assigned to a vertex



Evaluating impact of sensor degradation

Reconstruction performance at 0 fb⁻¹



Reconstruction performance at 1000 fb⁻¹



Reconstruction performance at 4000 fb⁻¹



Outlook

- We managed to include HGTD time information in the CKF track reconstruction
 - Still need to evaluate (and improve) track efficiency (second part of AQP)
- This reconstruction chain can be used to assert sensor performance in different phases of its lifetime
 - Need to get more accurate degradation info with sensor group

Tracking challenges on HL-LHC

- Today's track reconstruction (*tracking*) process will not be fast enough to operate in a high luminosity scenario
- For Run 3, ATLAS utilizes Combinatorial Kalman Filter (CKF) for tracking
 - O(N²) processing time, N being the number of vertices (depositions)



Additional plots on sensor degradation

Reconstruction performance at 1001 fb⁻¹



Reconstruction performance at 2000 fb⁻¹



Reconstruction performance at 2001 fb⁻¹

