

Using Time Information in ACTS Vertexing Algorithms

Ideas & Concepts

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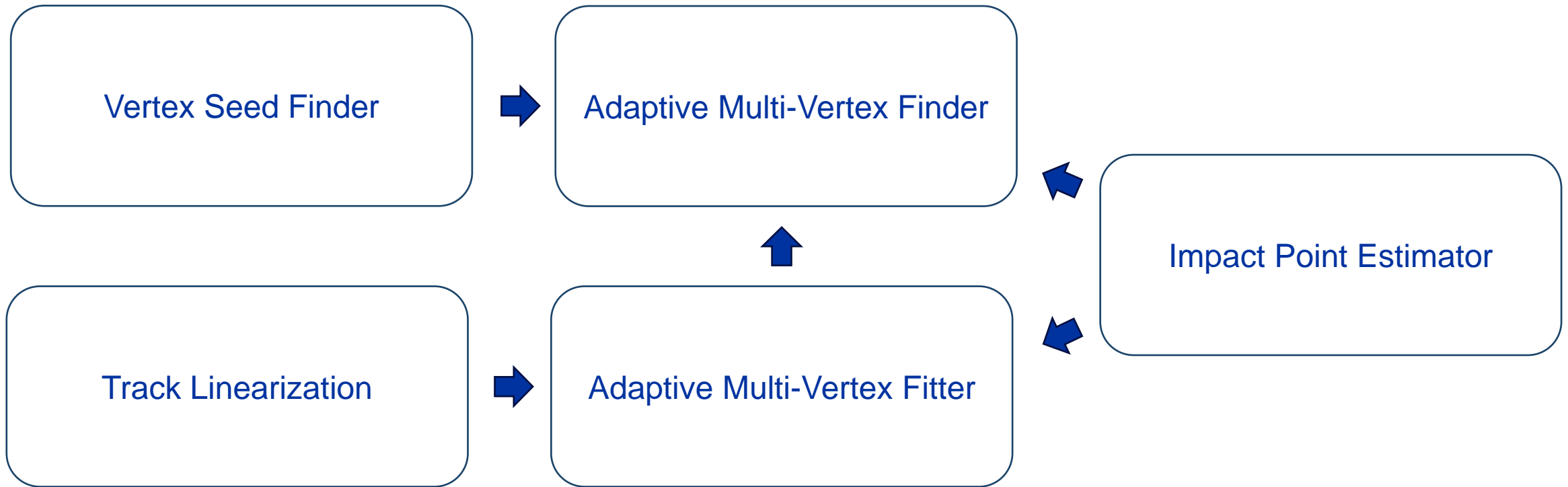
Task



$$t \in T \sim \hat{\mathbf{q}} = \left(d_0, z_0, \phi, \theta, \frac{q}{p}, t_0 \right)^1$$

$$v \in V \sim (x, y, z, t) \\ \text{and} \\ \mathbf{p} \forall t \in T_v \subset T$$

Simplified Structure



Vertex Seed Finding

AdaptiveGridTrackDensity.*pp & AdaptiveGridDensityVertexFinder.*pp

- Goal: Find a first estimate of the vertex position from a set of tracks
- Tracks are modeled as a 2D Gaussian distribution in the d - z -plane¹
- Density at $d = 0$ is calculated at discrete z -values for each track and added to a map

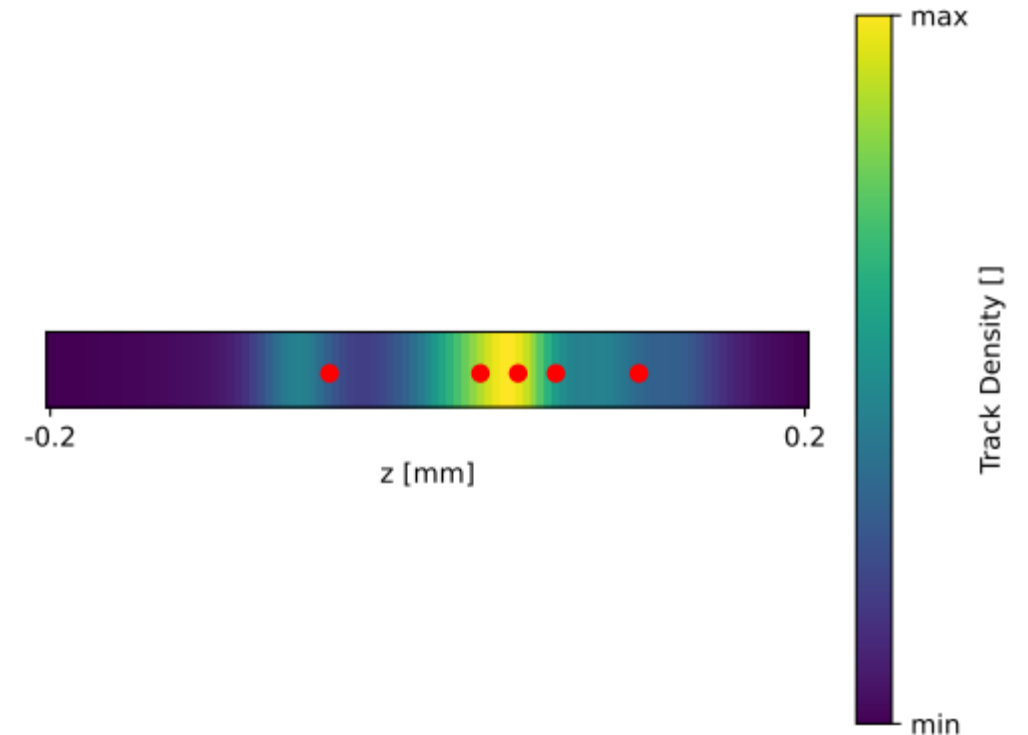
$M: \text{bin} \rightarrow \text{densityValue}$

- Cache track densities \rightarrow can be removed without recalculating their contribution
- We effectively have a 1D density grid; its maxima are the vertex seeds $(0, 0, z_{\text{max}}, 0)$

Vertex Seed Finding

AdaptiveGridTrackDensity.*pp & AdaptiveGridDensityVertexFinder.*pp

- Data:
 - Five vertices (red dots) at random positions
 - 4 Muons per vertex
 - Tracks reconstructed using default seeding and Combinatorial Kalman Filter

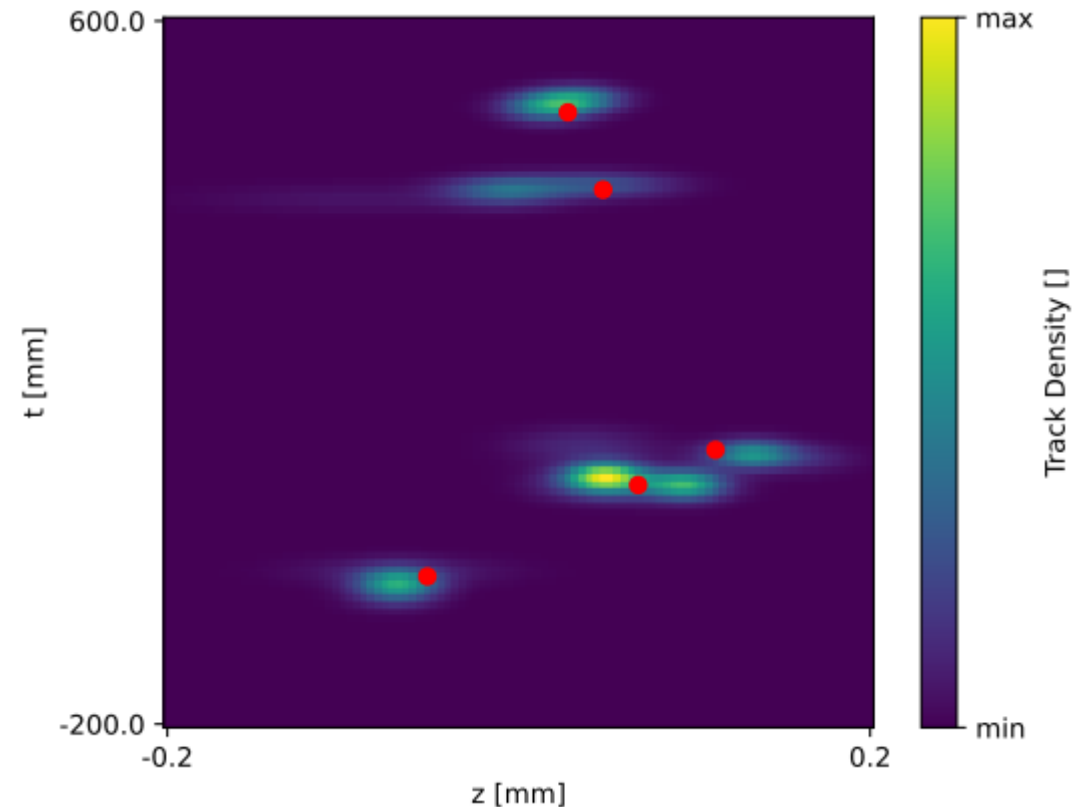


- In high-luminosity environments, the 1D description is not sufficient to resolve all vertices¹

Vertex Seed Finding

AdaptiveGridTrackDensity.*pp & AdaptiveGridDensityVertexFinder.*pp

- Time can be included using a 3D distribution in the d - z - t -plane
- As before, we evaluate at $d = 0$ to obtain a 2D grid
- Its maxima are the vertex seeds $(0, 0, z_{\max}, t_{\max})$
- Vertices can be resolved now!



Impact Point Estimation

ImpactPointEstimator.*pp

- Goal: Estimate compatibility between vertices and tracks
- Find the 3D point of closest approach (PCA) between the track and the vertex
 - Use the Newton method to find a minimum of the distance
- Propagate to the PCA, i.e., a plane reference surface with its origin at the vertex
 - x -axis in direction of the PCA
 - z -axis in direction of the momentum
 - y -axis follows from orthogonality

Impact Point Estimation

ImpactPointEstimator.*pp

- Calculate $\chi_{\text{vtx,trk}}^2 = \bar{\mathbf{r}}^T \mathbf{K}_{\mathbf{r}_{\text{trk}}}^{-1} \bar{\mathbf{r}}$, where:

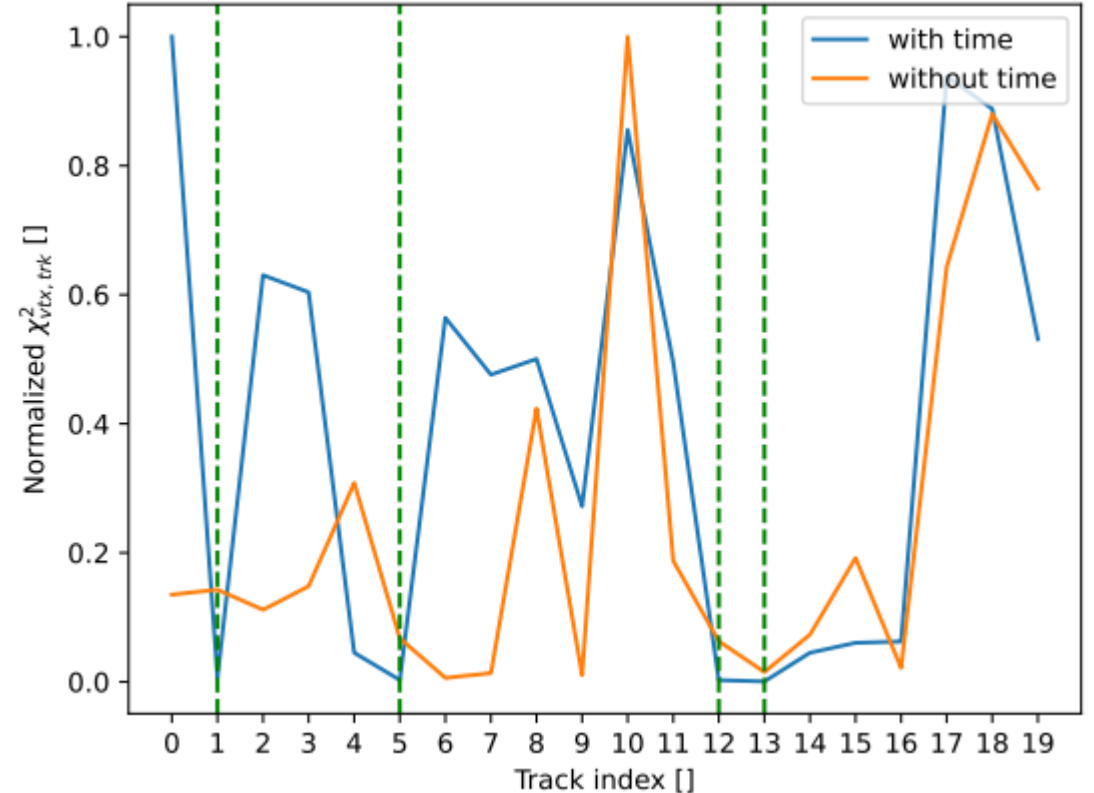
$$\begin{aligned} - \bar{\mathbf{r}} &= \mathbf{r}_{\text{trk}} - \mathbf{r}_{\text{vtx}} & - \mathbf{r}_{\text{trk}} &= \begin{pmatrix} x_{\text{trk}} \\ y_{\text{trk}} \\ t_{\text{trk}} \end{pmatrix} & - \mathbf{r}_{\text{vtx}} &= \begin{pmatrix} 0 \\ 0 \\ t_{\text{vtx}} \end{pmatrix} \end{aligned}$$

- Note that $z_{\text{trk}} = z_{\text{vtx}} = 0$ because both are on the reference surface

Impact Point Estimation

ImpactPointEstimator.*pp

- Same data as before
- Choose a random vertex
- Calculate $\chi^2_{\text{vtx}, \text{trk}}$ for all tracks with and without time
- The green lines indicate the tracks that really originate at the vertex



Adaptive Multi-Vertex Fitter

AdaptiveMultiVertexFitter.*pp & KalmanVertexUpdater.*pp,

- Based on a Kalman Filtering approach + annealing
- Minimizes $\chi_{total}^2 = \sum_{\text{vertices}} \sum_{\text{tracks}} (w(T, \chi_{\text{vtx}, \text{trk}}^2) \bar{\mathbf{q}}^T \mathbf{K}_{\hat{\mathbf{q}}}^{-1} \bar{\mathbf{q}})$
- $\bar{\mathbf{q}} = \mathbf{q}_{\text{model}} - \hat{\mathbf{q}}$
- $\hat{\mathbf{q}}$ and $\mathbf{K}_{\hat{\mathbf{q}}}^{-1}$ come from the tracking; $\chi_{\text{vtx}, \text{trk}}^2$ comes from the impact point estimation
- $\mathbf{q}_{\text{model}}(\mathbf{r}_V, \mathbf{p}_V) \approx \mathbf{q}(\mathbf{r}_{\text{PCA}}, \mathbf{p}_{\text{PCA}}) + \mathbf{A}(\mathbf{r}_V - \mathbf{r}_{\text{PCA}}) + \mathbf{B}(\mathbf{p}_V - \mathbf{p}_{\text{PCA}})$
- $\mathbf{A} = \frac{\partial \mathbf{q}}{\partial \mathbf{r}_W} |_{W=\text{PCA}}$ and $\mathbf{B} = \frac{\partial \mathbf{q}}{\partial \mathbf{p}_W} |_{W=\text{PCA}}$ are the Jacobians

Track Linearization

HelicalTrackLinearizer.*pp

- Analytically:

$$\begin{array}{cc}
 \left. \frac{d\mathbf{q}}{dx_w} \right|_{W=PCA} & \left. \frac{d\mathbf{q}}{dy_w} \right|_{W=PCA} & \left. \frac{d\mathbf{q}}{dz_w} \right|_{W=PCA} & \left. \frac{d\mathbf{q}}{dt_w} \right|_{W=PCA} \\
 \downarrow & \downarrow & \downarrow & \downarrow \\
 \mathbf{A} = \begin{pmatrix}
 -\sin(\phi) & \cos(\phi) & 0 & 0 \\
 -\frac{|\rho|}{S} \cot(\theta) \cos(\phi) & -\frac{|\rho|}{S} \cot(\theta) \sin(\phi) & 1 & 0 \\
 -\frac{\text{sgn}(\rho)}{S} \cos(\phi) & -\frac{\text{sgn}(\rho)}{S} \cos(\phi) & 0 & 0 \\
 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 \\
 -\frac{|\rho|}{c\beta_T S} \cos(\phi) & -\frac{|\rho|}{c\beta_T S} \sin(\phi) & 0 & 1
 \end{pmatrix}
 \end{array}$$

- ρ ... signed helix radius
- S ... $x - y$ - distance between vertex and helix center
- $c\beta_T$... speed in the $x - y$ - plane

Track Linearization

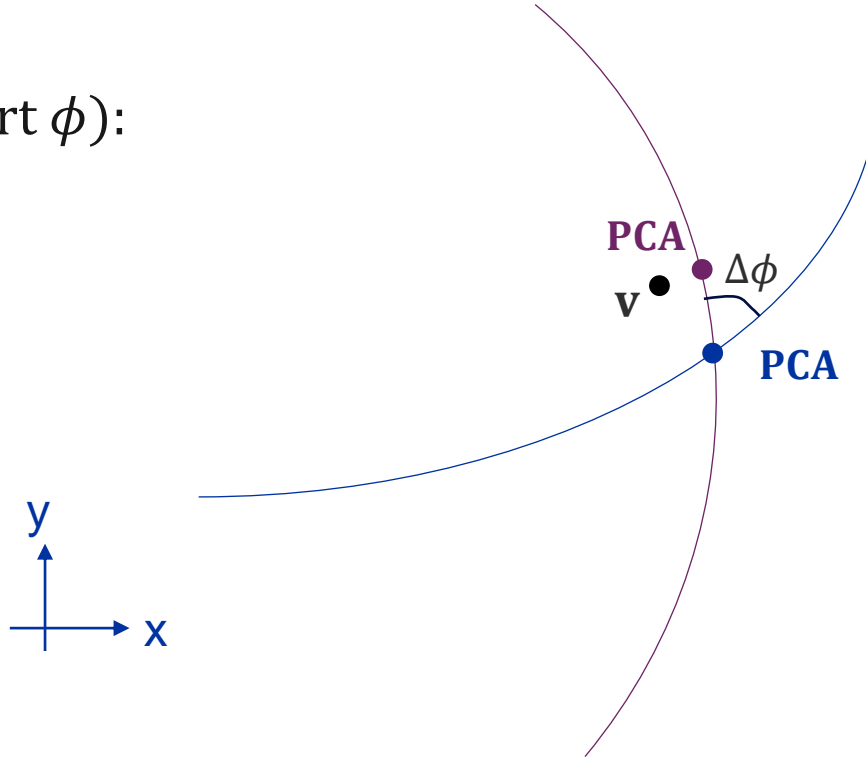
HelicalTrackLinearizer.*pp

- Assumes a constant magnetic field
- For the other Jacobian \mathbf{B} , also the last row had to be added (not shown here)
- Derived together with P. Butti – thanks a lot!
- Results checked numerically

Track Linearization

NumericalTrackLinearizer.*pp

- Numerically (here for derivatives wrt ϕ):



- $\frac{\partial \mathbf{q}}{\partial \phi} \Big|_{\text{PCA}} \approx \frac{\hat{\mathbf{q}}(\phi + \Delta\phi) - \hat{\mathbf{q}}(\phi)}{\Delta\phi}$
- Computationally expensive, but works for a non-constant magnetic field
→ Potentially useful for secondary vertexing

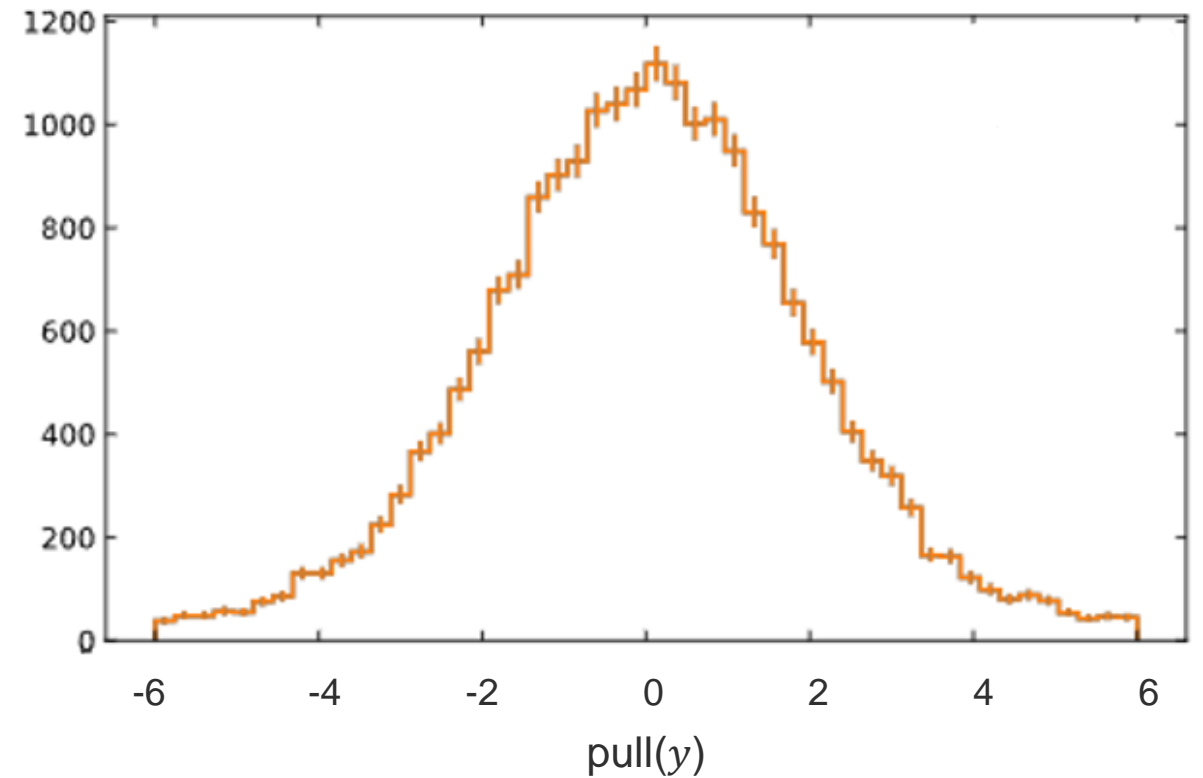
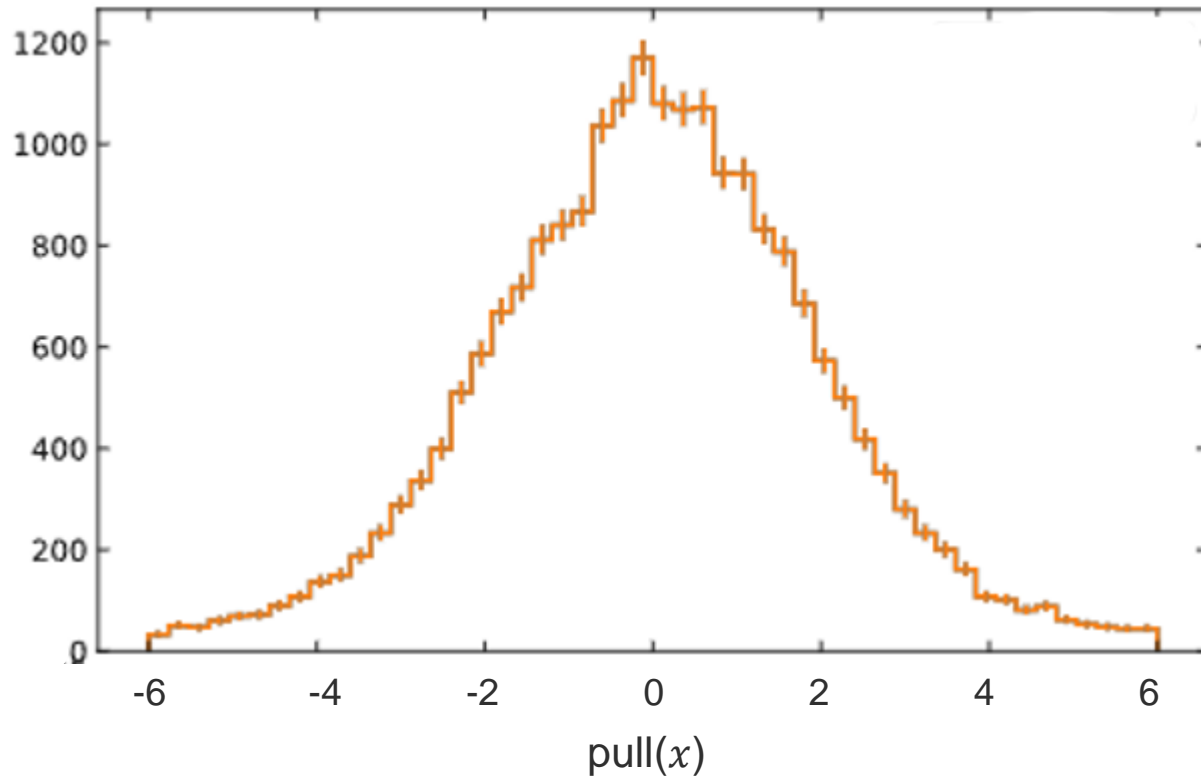
Results from Billoir Vertex Fit

FullBilloirVertexFitter.*pp, physmon_ckf_tracking.py

- Fitting the vertex time using the Billoir method
 - Mathematically equivalent to the Kalman vertex fit
 - Kalman vertex fitter is not ready yet
- Using 500 4-Muon events
- Pseudo pile-up of 50
 - Particle gun with vertex smearing
- Default seeding + Combinatorial Kalman Filter

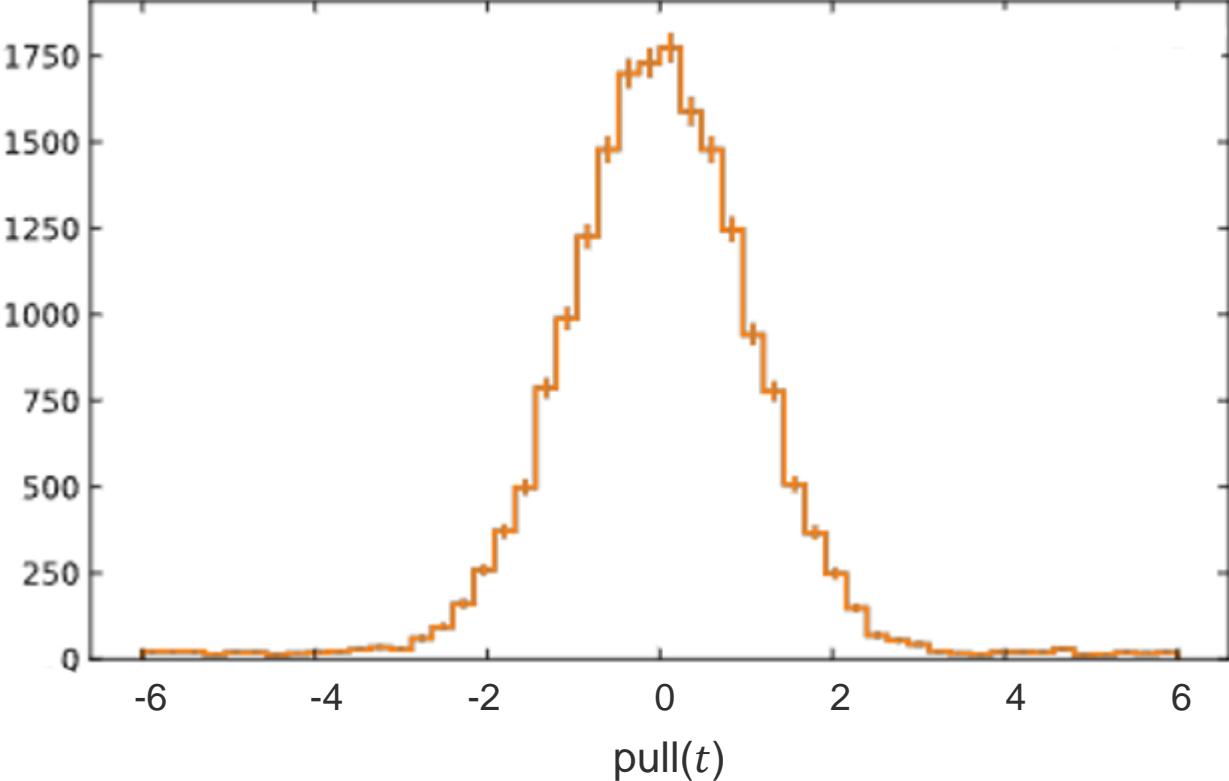
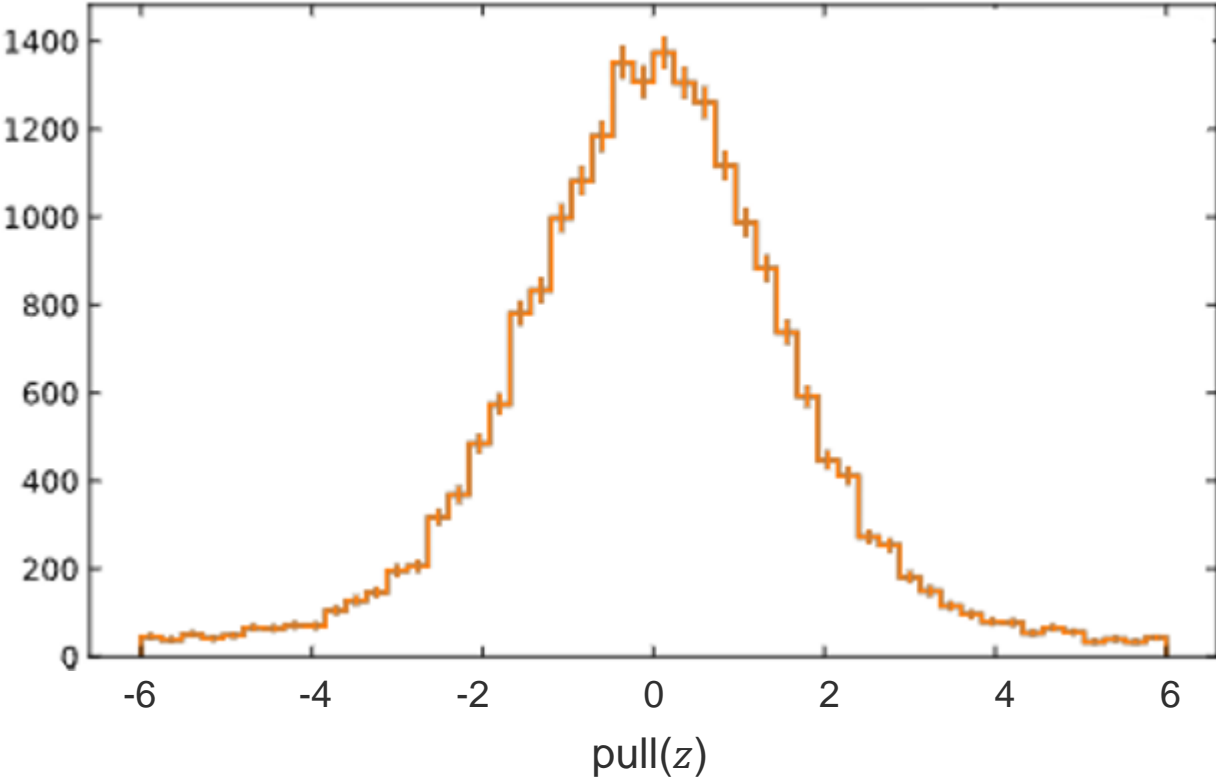
Results from Billoir Vertex Fit

FullBilloirVertexFitter.*pp, physmon_ckf_tracking.py



Results from Billoir Vertex Fit

FullBilloirVertexFitter.*pp, physmon_ckf_tracking.py



Conclusion

- Time is gradually introduced into the ACTS vertexing suite
- Can be included elegantly by adding a new dimension
- Partial time measurements should not be a problem
 - If a tracks does not have a time measurement, we inflate the corresponding covariance
- Time vertexing is promising enhanced resolution in high-luminosity environments
- Big thanks to: P. Butti, P. Gessinger-Befurt, A. Salzburger, B. Schlag, A. Stefl

Backup slides

Vertex Seed Finding

AdaptiveGridTrackDensity.*pp & AdaptiveGridDensityVertexFinder.*pp

- Mathematical model

$$P(d, z) \propto \frac{1}{\det(\mathbf{K}_{\widehat{\mathbf{q}}_{\text{IP}}})} \exp(\overline{\mathbf{q}}_{\text{IP}}^T \mathbf{K}_{\widehat{\mathbf{q}}_{\text{IP}}}^{-1} \overline{\mathbf{q}}_{\text{IP}})$$

$$\mathbf{q}_{\text{IP}} = \mathbf{q}_{\text{IP}} - \widehat{\mathbf{q}}_{\text{IP}} \qquad \mathbf{q}_{\text{IP}} = \begin{pmatrix} d \\ z \end{pmatrix} \qquad \widehat{\mathbf{q}}_{\text{IP}} = \begin{pmatrix} d_0 \\ z_0 \end{pmatrix}$$

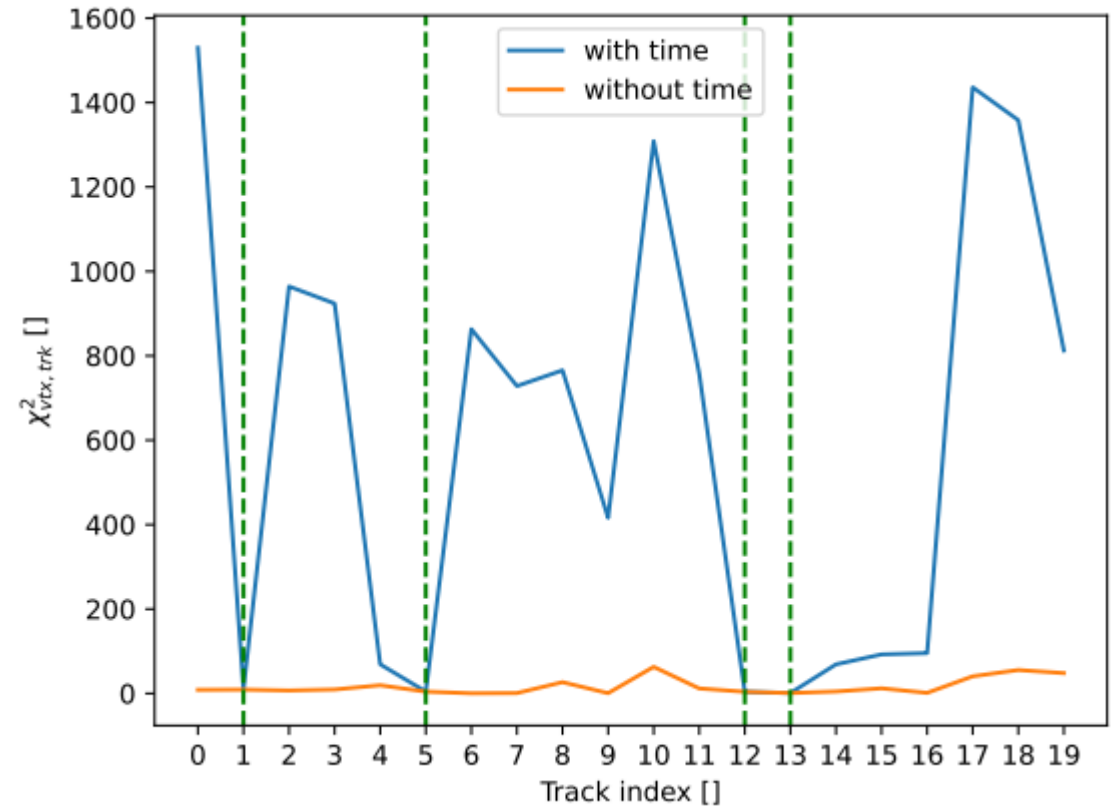
- Using time:

$$\mathbf{q}_{\text{IP}} = \begin{pmatrix} d \\ z \\ t \end{pmatrix} \qquad \widehat{\mathbf{q}}_{\text{IP}} = \begin{pmatrix} d_0 \\ z_0 \\ t_0 \end{pmatrix}$$

Impact Point Estimation

ImpactPointEstimator.*pp

- Unnormalized $\chi^2_{\text{vtx, trk}}$



Adaptive Multi-Vertex Fitter

AdaptiveMultiVertexFitter.*pp & KalmanVertexUpdater.*pp,

- Based on a Kalman Filtering approach + annealing
- Adding a new measurement ~ adding a track to the fit
- Temperature is decreased \rightarrow outliers are weighted down gradually
- $\chi^2 = \chi_{\text{vtx, trk}}^2$ from impact point estimation
- A track can be associated to multiple vertices
 - Vertices are competing for tracks

