



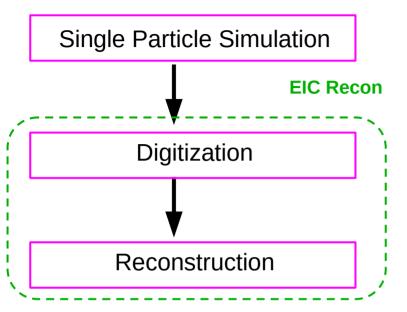
Benchmarking Tracking Performance and Charm Hadron Reconstruction in the ePIC Experiment

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Previous studies (Average number of hits)

 $https://indico.bnl.gov/event/23130/contributions/90774/attachments/54104/92571/Tracking_Performances_Shyam_25April24.pdf$

Simulation and Reconstruction in ePIC experiment

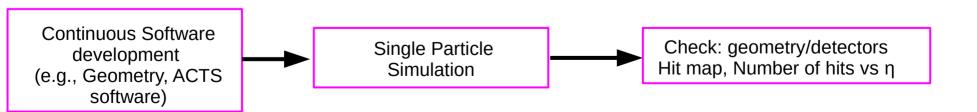


DD4HEP and GEANT4 (Particle Gun): Single particle generation and their propagation using G4ParticleGun in GEANT4

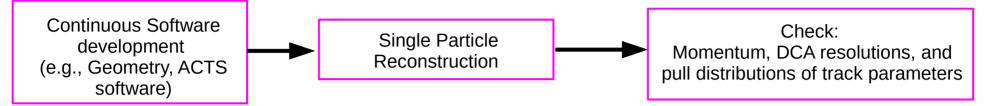
Digitization (detector response): For silicon detectors hits are smeared (Gaussian) according to spatial resolutions after applying an energy threshold

Tracking in EIC Recon using ACTS (A Common Tracking Software): Digitized hits are used for the track reconstruction in ACTS framework

Why single particle tracking benchmarks?



Simulation and Reconstruction in ePIC experiment



Two main options:

- → Test mode: run 10k events
- Full simulation campaigns

GNU: Makefile (compile) Python: Snakemake

Snakefile setting: <u>Snakefile</u>

Concept is similar: target-rule (command)

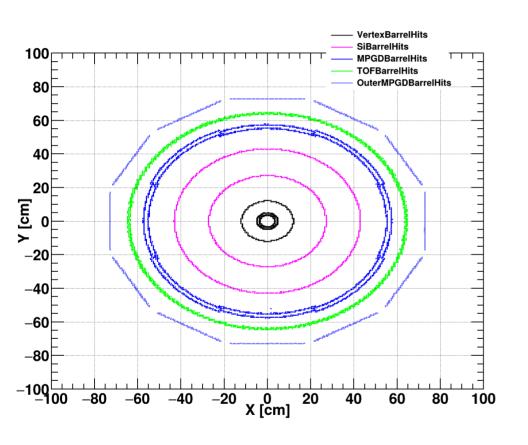
Tracking performances was not working fine: Fixed

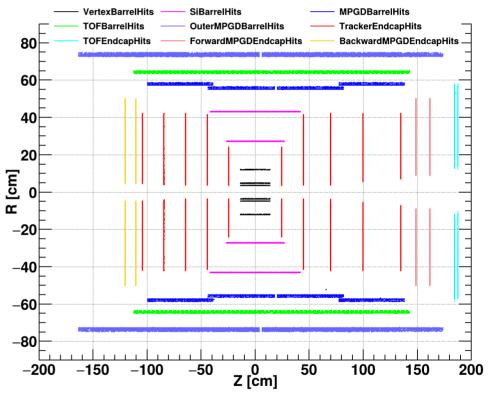
```
@@ -77,9 +77,9 @@ void Tracking_Performances(TString filename="tracking_output",TString particle="
     TTreeReaderArray<Double t> vx mc(mvReader, "MCParticles.vertex.x");
     TTreeReaderArray<Double_t> vy_mc(myReader, "MCParticles.vertex.y");
     TTreeReaderArray<Float_t> px_mc(myReader, "MCParticles.momentum.x");
     TTreeReaderArray<Float_t> py_mc(myReader, "MCParticles.momentum.y");
     TTreeReaderArray<Float_t> pz_mc(myReader, "MCParticles.momentum.z");
     TTreeReaderArray<Int_t> status(myReader, "MCParticles.generatorStatus");
     TTreeReaderArray<Int_t> pdg(myReader, "MCParticles.PDG");
     TTreeReaderArray<Int t> match flag(myReader,
  Form("CentralCKF%sTrackParameters.type", tag.Data()));
     TTreeReaderArray<Double_t> vx_mc(myReader, "MCParticles.vertex.x");
      TTreeReaderArray<Double_t> vy_mc(myReader, "MCParticles.vertex.y");
                                                                                Change type from
                                                                                  float to double
      TTreeReaderArray<Double t> px mc(myReader, "MCParticles.momentum.x");
      TTreeReaderArray<Double t> pv mc(mvReader, "MCParticles.momentum.v");
     TTreeReaderArray<Int_t> pdg(myReader, "MCParticles.PDG");
     TTreeReaderArray<Int_t> match_flag(myReader
   Form("CentralCKF%sTrackParameters.type",tag.Data()));
```

Hit Maps (Simulation Level)

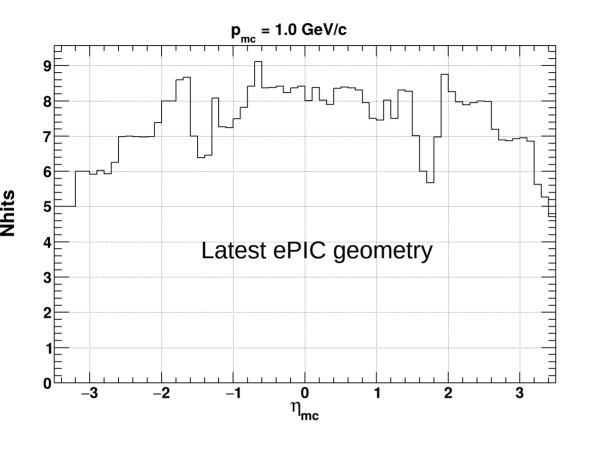
10k Events (pi-)

MOMENTUM=[0.5, 1.0, 2.0, 5.0, 10.0, 15.0]





Average Number of Hits vs η



Nhits = 3 will lead to biased results due to the low number of degrees of freedom (ndf = 1)

Similar plots for other momentum ranges are produced

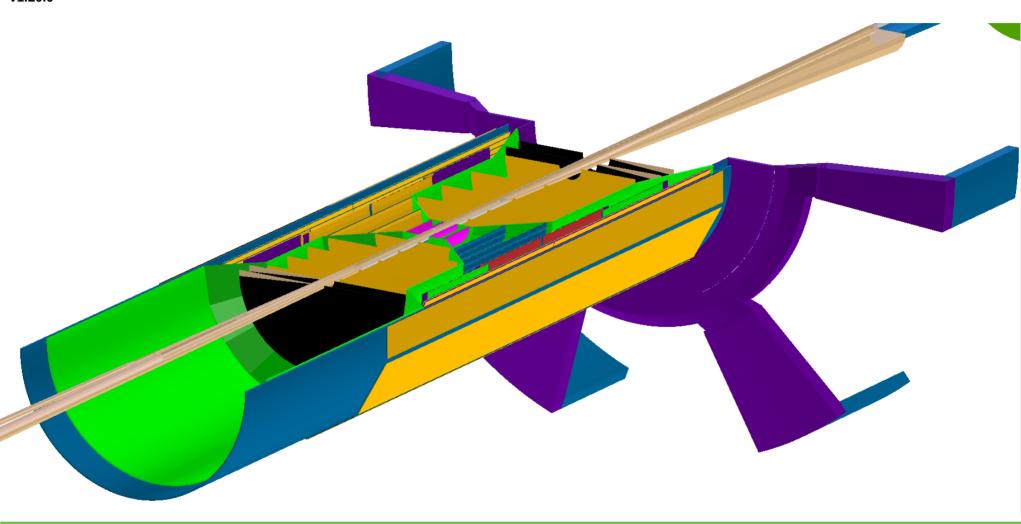
Tracking Performances Results

https://github.com/eic/detector_benchmarks/pull/113

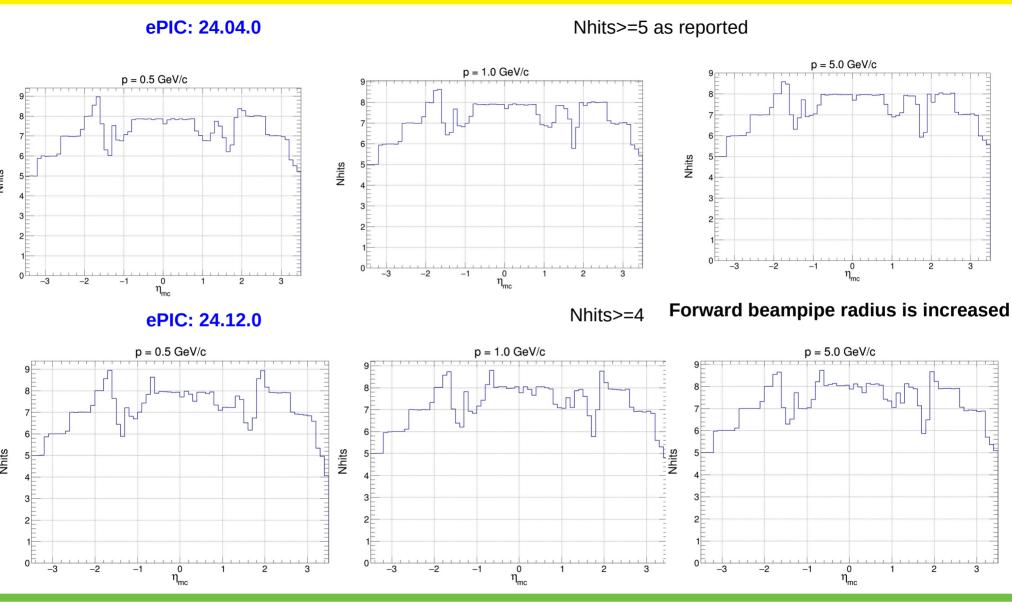
The ePIC Tracker

shyam@shyam:~/eic/epic\$ git describe --tags --abbrev=0 **24.12.0** shyam@shyam:~/eic/EICrecon\$ git describe --tags --abbrev=0 **v1.20.0**

Geometry Parameters

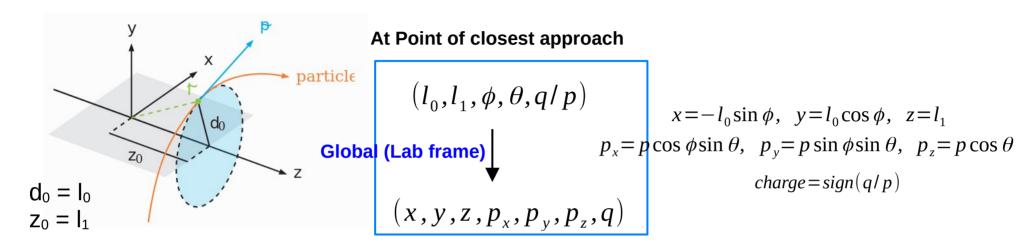


Nhits vs Eta



Track Parameters in ACTS

Track Parameters $(l_0, l_1, \phi, \theta, q/p, time)$ If tracking algorithm is working fine Pull must be consistent with unity



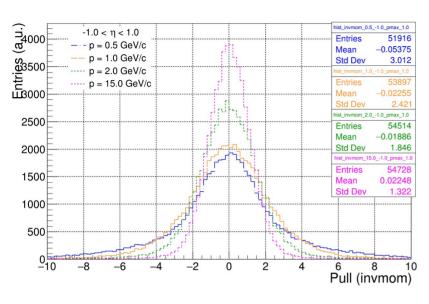
Plan to add it to the benchmarks

$$\begin{aligned} & \text{Pull } l_0 = \frac{(l_{0\,\text{rec}} - l_{0\,\text{gen}})}{\sigma_{l0}} & \text{Pull } \phi = \frac{(\phi_{\text{rec}} - \phi_{\text{gen}})}{\sigma_{\phi}} & \text{Pull } q/p = \frac{(q/p_{\text{rec}} - q/p_{\text{gen}})}{\sigma_{q/p}} \\ & \text{Pull } l_1 = \frac{(l_{1\,\text{rec}} - l_{1\,\text{gen}})}{\sigma_{l1}} & \text{Pull } \theta = \frac{(\phi_{\text{rec}} - \phi_{\text{gen}})}{\sigma_{\theta}} & \text{Pull } p = \frac{(p_{\text{rec}} - p_{\text{gen}})}{\sigma_{p}} \end{aligned}$$

Simulation of 200K pi- for momentum 0.5, 1.0, 2,0, 15.0 GeV/c

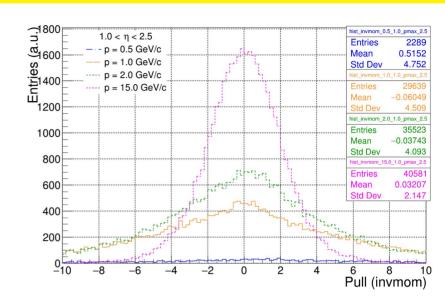
ePIC: 24.12.0 EICRecon: v1.20.0

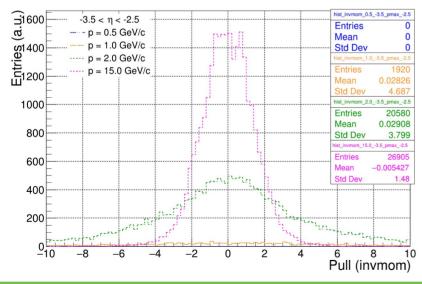
Pull distributions (Inverse Momentum)



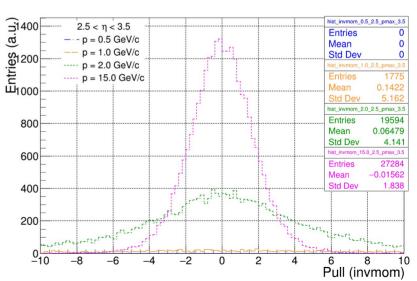
ePIC: 24.12.0

EICRecon: v1.20.0

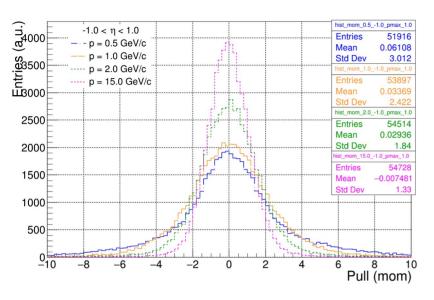


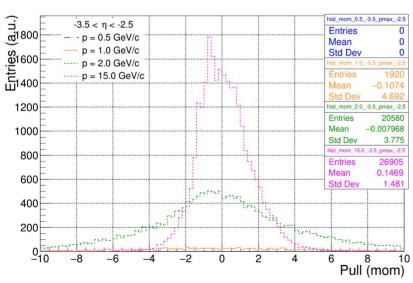


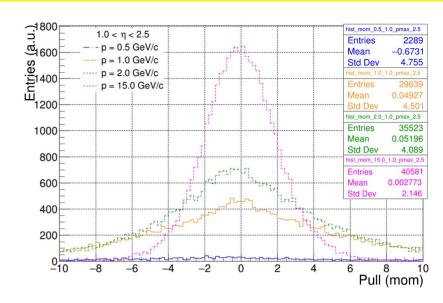
Inverse Momentum = 1/p

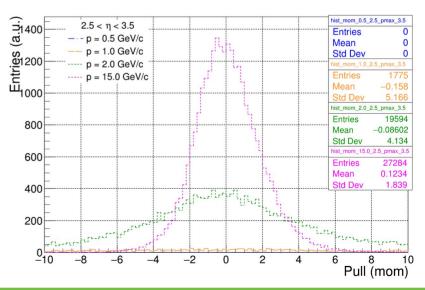


Pull distributions (Momentum)

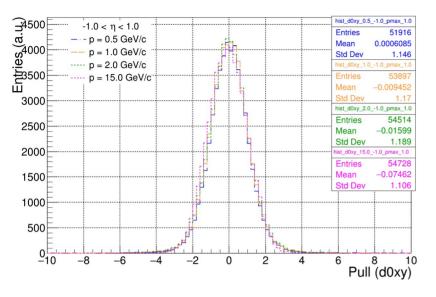


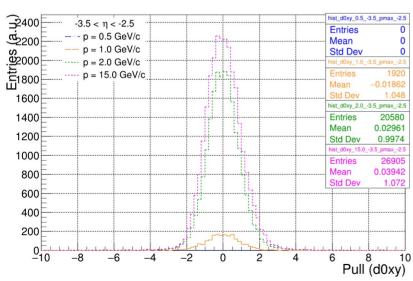


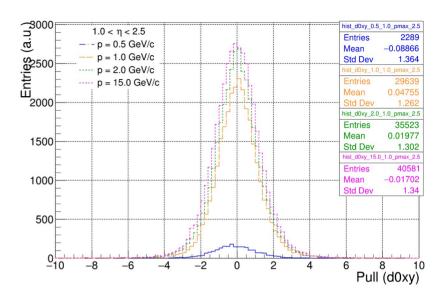


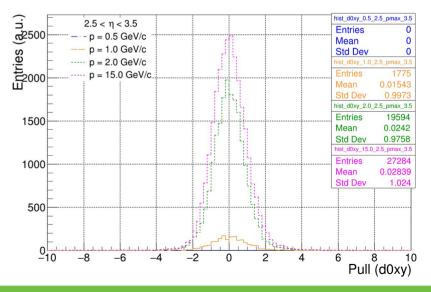


Pull distributions (I_0 or d_{0xy})

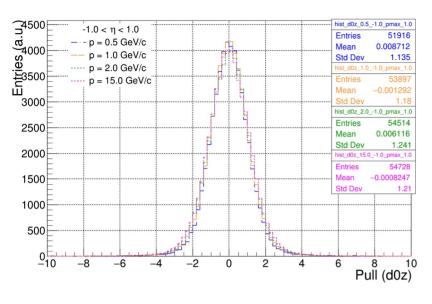


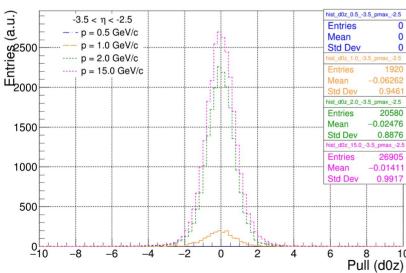


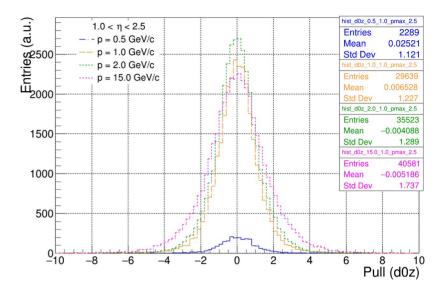


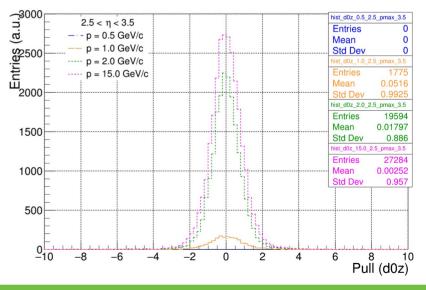


Pull distributions (I₁ or d₀₇)

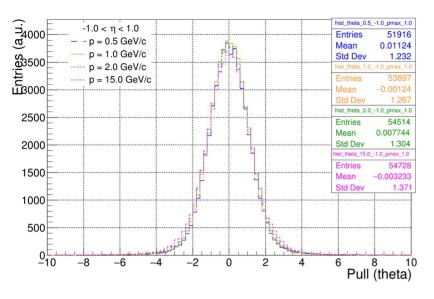


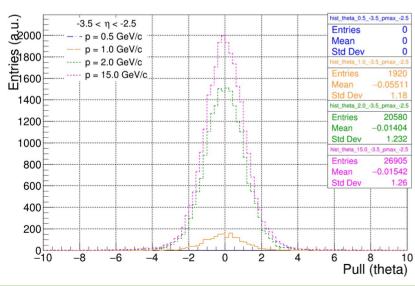


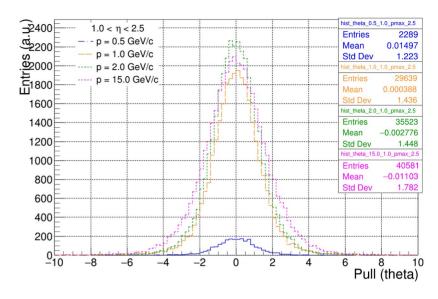


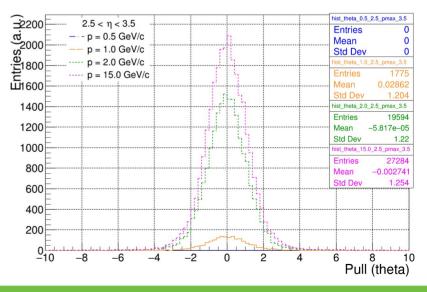


Pull distributions (Theta)

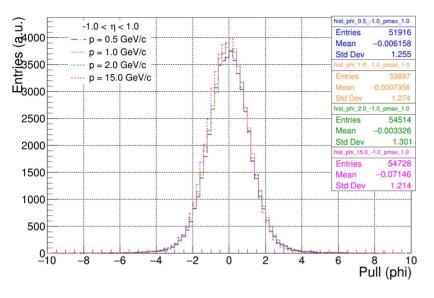


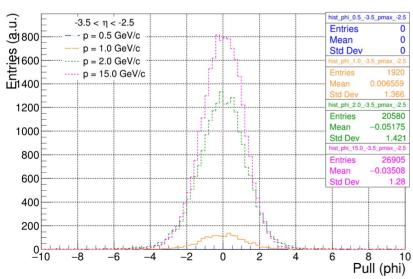


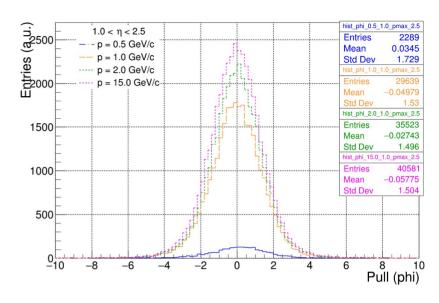


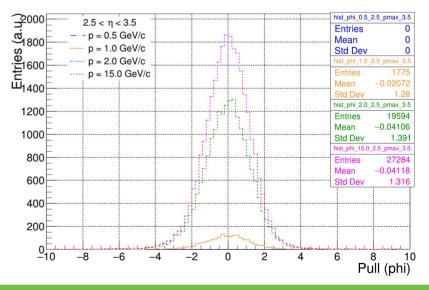


Pull distributions (Phi)

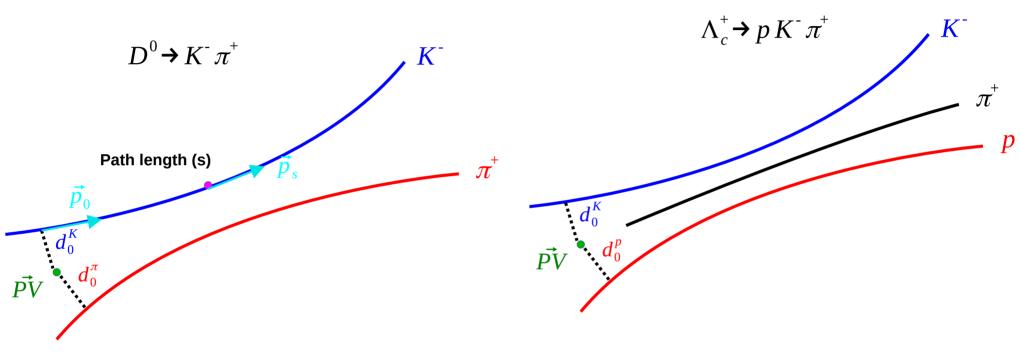






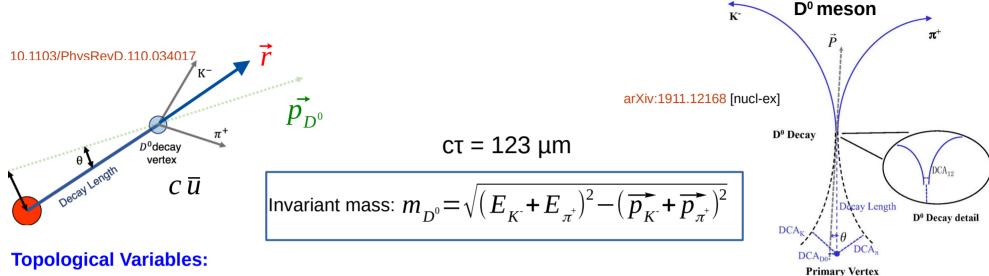


Charm Hadron Reconstruction



Slides from Jets and Heavy-flavor meeting

Topological Variables



- DCA_k and DCA_n with respect to the reconstructed primary vertex (d0 k, d0 pi)
- Decay length of D⁰ meson (decaylength)
- $Cos\theta$ (angle between **dl** and **p**_{D0})
- DCA₁₂ distance between the daughter tracks of D⁰
- DCA_{D0} impact parameter of reconstructed D⁰ meson
- m_{D0} invariant mass of kaon and pion pairs
- pt D0 reconstructed pt of the D0 meson
- eta D0 reconstructed η of the D⁰ meson
- Multiplicity (mult)

Decay length (dl), Primary Vertex (PV), Secondary Vertex (SV)

$$\vec{dl} = \vec{SV} - \vec{PV}$$

$$\cos \theta = \frac{\vec{dl} \cdot \vec{p}}{|\vec{dl}||\vec{p}|}$$

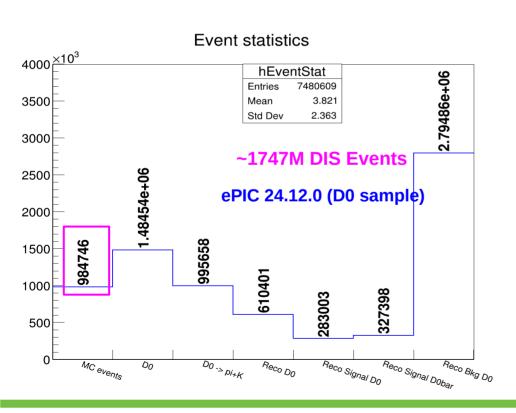
$$DCA_{D0} = dl \sin \theta$$

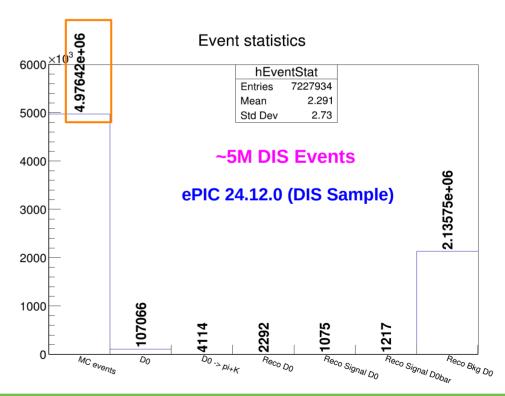
Data Sample for ML ($Q^2 > 1 \text{ GeV}^2$)

- \triangleright BDT requires the features for the signal D $^{\circ}$ meson and background D $^{\circ}$ meson (fake combinations of pion,kaon)
 - D⁰ enriched same created filtering PYTHIA8 ep, NC, 10X100, Q² >1 GeV² events (~1747M) such that each event consist one D⁰ → k-π+ known as Signal taken from 24.12.0/epic_craterlake/SIDIS/D0_ABCONV/pythia8.306-1.1/10x100/q2_1):

Total files 1879 and Events = 984746

Background from 24.12.0/epic_craterlake/DIS/NC/10x100/minQ2=1: Total files 5180 and Events = 4976419





Secondary Vertex Reconstruction (D⁰)

Secondary Vertex

Approach 1 (Analytical)

$$\vec{SV} = \frac{\vec{pca}_1 + \vec{pca}_2}{2}$$

Secondary vertexing in ACTS considers tracking errors properly

 \vec{PV}

Ignored track errors (at the moment)

$$Track_{DCA} = (\vec{r_0}, \vec{p_0}, q)$$

 $Track_{At(s)} = (\vec{r}_s, \vec{p}_s, q)$ s: path length

Comparison of four approaches:

- → Helix1 (using helix1 to find pca₁ and pca₂)
- Distance minimization (d)
- → Helix2 (using helix2 to find pca₁ and pca₂)
- → Using average of Helix1 and Helix2

$D^0 \rightarrow K^- \pi^+$

 $DCA_{\pi K} = |\vec{pca_1} - \vec{pca_2}|$

pca SV

pca₂

Approach 2 (Shyam) Minimizing the distance

Vertex position $\overrightarrow{SV} = (v_x, v_y, v_z)$

Total parameters (5) = $(v_x, v_y, v_z, s_1, s_2)$

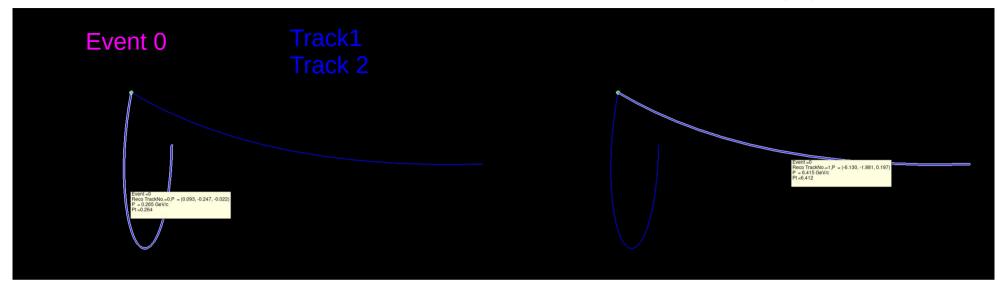
$$Track_{At(s1)} = (\vec{r_{s1}}, \vec{p_{s1}}, q_1)$$

 $Track_{At(s2)} = (\vec{r_{s2}}, \vec{p_{s2}}, q_2)$

Minimize $d = \sqrt{(\vec{r}_{s1} - \vec{v})^2 + (\vec{r}_{s2} - \vec{v})^2}$

Event display (First two tracks)

Root Based: Event display



- Visualization of two tracks in Event 0 (Either Signal pair or Bkg pair)
- > Visualization of **pca** and secondary vertex with different approaches
- Distance minimization returns the unique point

Different approaches:

- → Helix1 (using helix1 to find pca₁ and pca₂)
- → Distance minimization (d)
- → Helix2 (using helix2 to find pca₁ and pca₂)

Event display (First two tracks)

Root Based: Event display

Event 0
Green marker track1
reference
Red marker track2

White marker (distance minimization)

reference

Green track1 All three methods (track 1 reference, track2 reference, and distance minimization) are compatible

Minor difference is due to analytical approach

Distance minimization returns the unique point

Checking overall performances for Signal and Bkg D0 meson (next slides) and few other events in backup

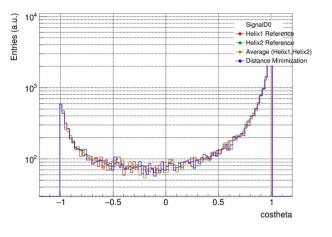
Comparison Distribution for D⁰ meson (Q²>1GeV²)

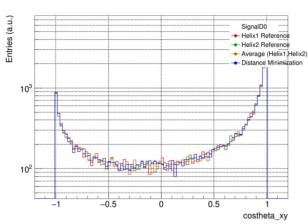
Signal D⁰ meson

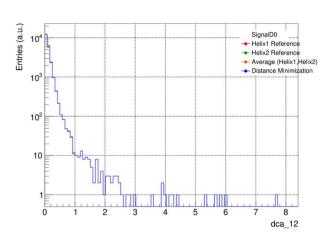
All methods are compatible

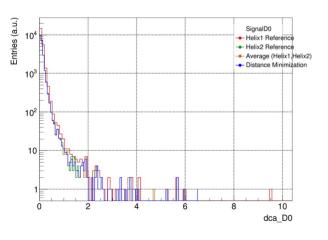


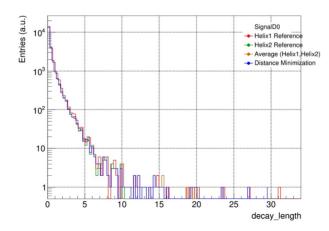
Distance minimization gives unique secondary vertex

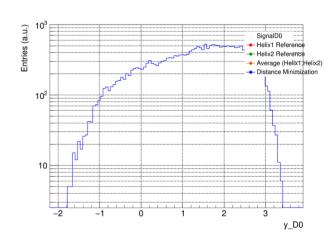








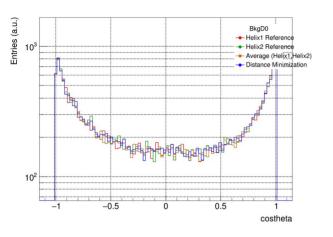


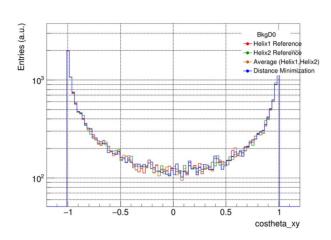


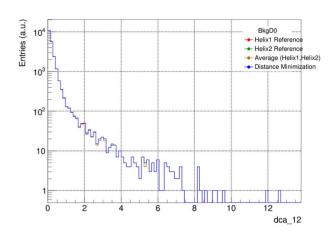
Comparison Distribution for D⁰ meson (Q²>1GeV²)

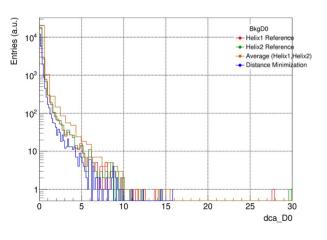
All methods are compatible

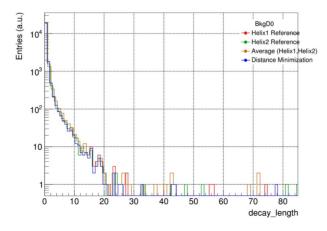
Bkg D⁰ meson

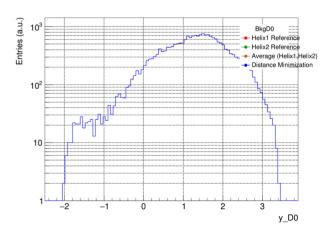












Secondary Vertex Reconstruction (Λ_{c}^{+})

Secondary Vertex

$$\Lambda_c^+ \rightarrow p K^- \pi^+$$

pca₁

 $p\vec{c}a_3$

 $DCA_{k\pi} = |p\vec{c}a_1 - p\vec{c}a_2|, \quad DCA_{kp} = |p\vec{c}a_1 - p\vec{c}a_3|, \quad DCA_{p\pi} = |p\vec{c}a_3 - p\vec{c}a_2|$

Vertex position $\overrightarrow{SV} = (v_x, v_y, v_z)$

 $DCA_{12} = min \{DCA_{nK}, DCA_{KP}, DCA_{nP}\} Cut$

Approach 1 (Analytical)

$$\vec{SV} = \frac{p\vec{c}a_1 + p\vec{c}a_2 + p\vec{c}a_3}{3}$$

Secondary vertexing in ACTS considers tracking errors properly

Ignored track errors (at the moment)

$$Track_{DCA} = (\vec{r_0}, \vec{p_0}, q).$$

$$Track_{At(s)} = (\vec{r}_s, \vec{p}_s, q)$$
 s: path length

 \vec{PV}

Total parameters (6) = $(v_x, v_y, v_z, s_1, s_2, s_3)$



Minimizing the distance

Total parameters (6) = $(v_x, v_y, v_z, s_1, s_2, s_3)$

$$Track_{At(s1)} = (\vec{r}_{s1}, \vec{p}_{s1}, q_1)$$

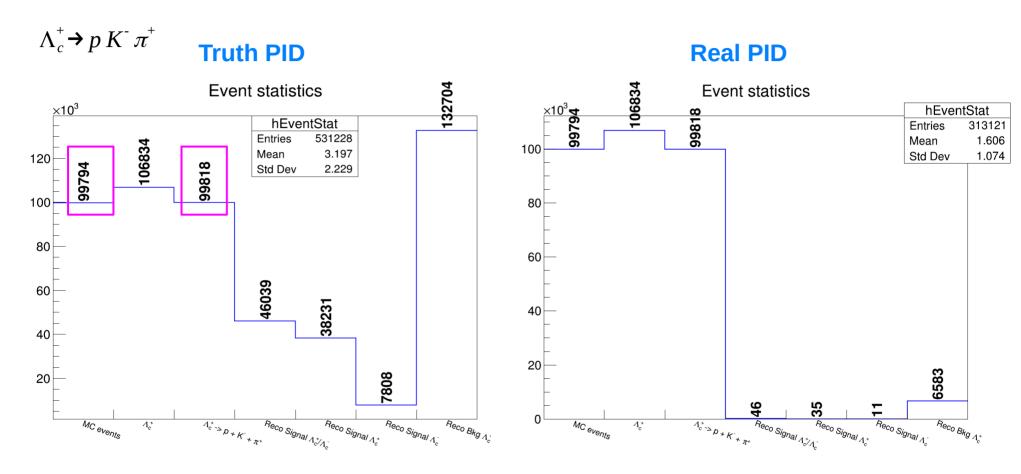
$$Track_{At(s2)} = (\vec{r_{s2}}, \vec{p_{s2}}, q_2)$$

$$Track_{At(s3)} = (\vec{r}_{s3}, \vec{p}_{s3}, q_3)$$

Minimize
$$d = \sqrt{(\vec{r_{s1}} - \vec{v})^2 + (\vec{r_{s2}} - \vec{v})^2 + (\vec{r_{s3}} - \vec{v})^2}$$

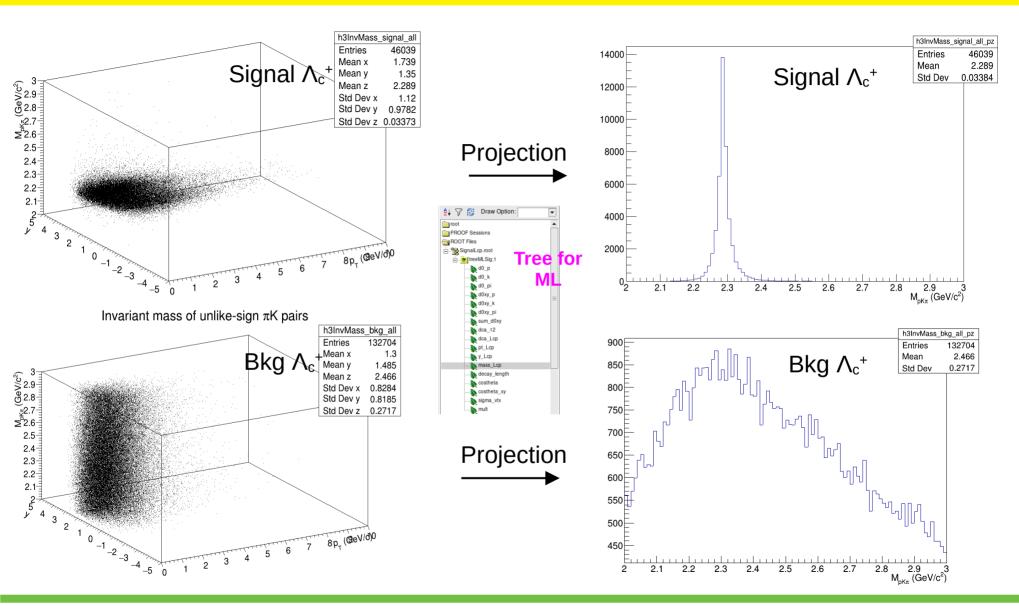
Λ_c⁺ Reconstruction

PYTHIA8 ep NC (10x100 Q²>1) Λ_c+ sample: by Rongrong

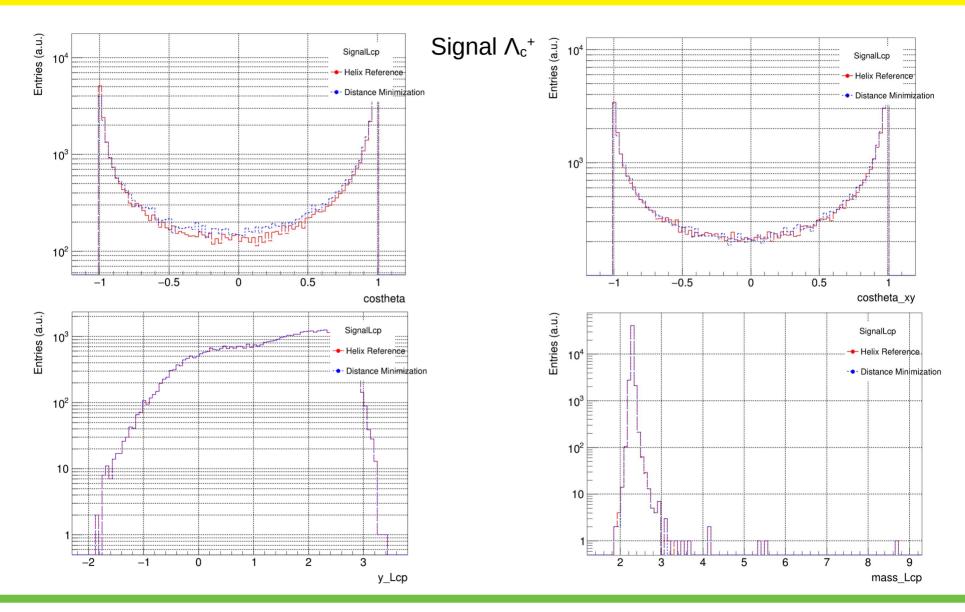


Issue coming from proton PID in Real PID, basically from lookup tables (loosing protons)

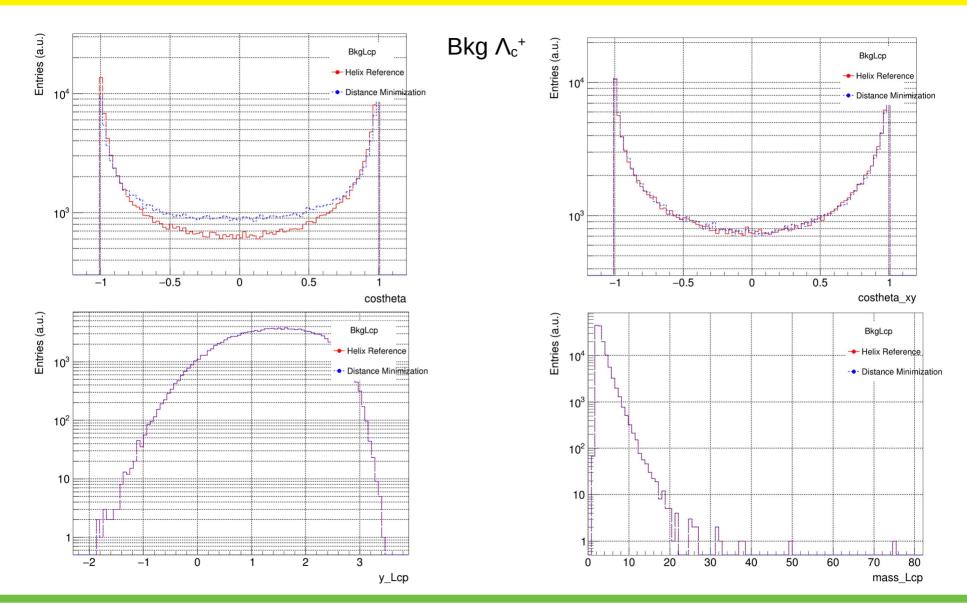
Results



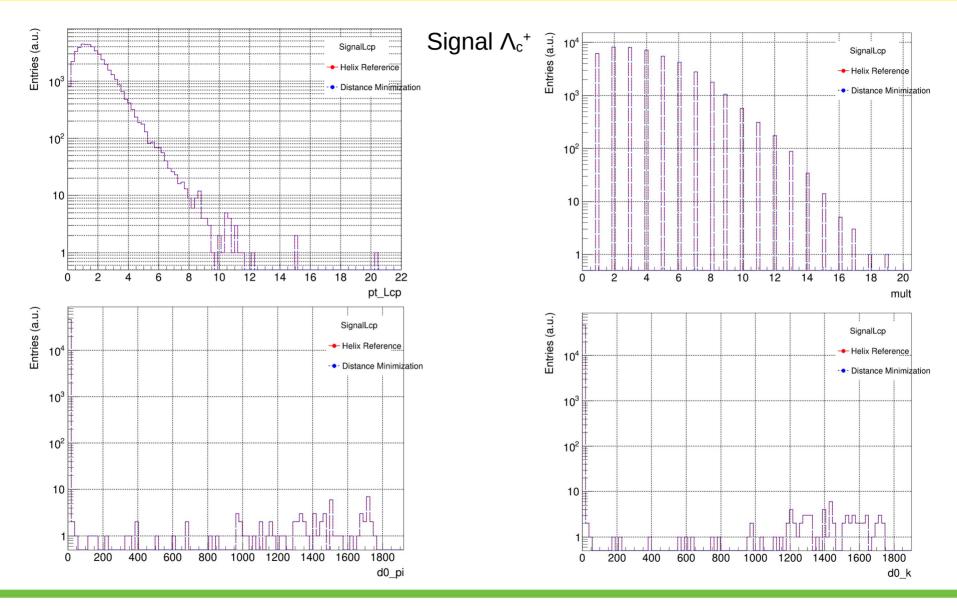
Comparison (Results)



Comparison (Results)



Comparison (Results)



Summary and Future Plan

- Tracking Performances are now working fine (Issue was related to data type conversion not the beampipe)
- Included hit maps and average number of hits per tracking to the benchmarks
- I further checked the pull distributions for the track parameters
- Compared with the different methods of extracting the secondary vertex, they are compatible with analytical method which is faster (we will use official code for secondary vertex in ACTS)
- \triangleright First implementation of Λ_c^+ reconstruction is completed and will commit the code soon
- Future Plan
 - Include hit maps and average number of hits per track for each momentum
 - Need to understand why pull distributions for momentum are not consistent with unity especially for higher η
 - Implement pull distributions to the benchmarks
 - Implementation of machine learning for Λ_c^+ is in progress

Track Parametrization (Local to Global)

Helical Track model:
$$(l_0, l_1, \phi, \theta, q/p)$$

 $x=-l_0\sin\phi, \ y=l_0\cos\phi, \ z=l_1$
 $p_x=p\cos\phi\sin\theta, \ p_y=p\sin\phi\sin\theta, \ p_z=p\cos\theta$
 $charge=sign(q/p)$

The particle $d_0=l_0$
 $z_0=l_1$

At Point of closest approach (perigee surface)

$$(l_0, l_1, \phi, \theta, q/p)$$

$$\downarrow \text{Global (Lab frame)}$$

$$(x, y, z, p_x, p_y, p_z, q)$$

```
Vector3 LineSurface::localToGlobal(const GeometryContext& gctx, const Vector2& lposition, const Vector3& direction) const

{

Vector3 unitZ0 = lineDirection(gctx);

// get the vector perpendicular to the momentum direction and the straw axis

Vector3 radiusAxisGlobal = unitZ0.cross(direction);

Vector3 locZinGlobal = transform(gctx) * Vector3(0., 0., lposition[1]);

// add loc0 * radiusAxis

return Vector3(locZinGlobal + lposition[0] * radiusAxisGlobal.normalized());

}

Calculation
```

UnitZ0: is (0,0,1) vector along the z-axis for cylinder and disks.

```
direction: (p Cos(phi) Sin(theta), p Sin(phi) Sin(theta), p Cos(theta))
radiusAxisGlobal = UnitZ0 Cross product direction = (-p Sin(phi) Sin(theta), p Cos(phi)
Sin(theta), 0)
radiusAxisGlobal.Normalized = (-Sin(phi), Cos(phi), 0)locZinGlobal = (0,0,11) (is same as global)
```

Global position = locZinGlobal + lposition[0] * radiusAxisGlobal.normalized() = (0,0,11) + <math>lo(-Sin(phi), Cos(phi), 0)Global Position = (-lo Sin(phi), lo Cos(phi), l1)

Returns the components, which we are using in HF analysis.

$$x = -I_0 \operatorname{Sin\phi}, y = I_0 \operatorname{Cos\phi}, z = I_1$$

Covariance Matrix in ACTS

Track Parameters $(l_0, l_1, \phi, \theta, q/p, time)$

For each fitted track we get track parameters and covariance matrix

Symmetric matrix: Independent entries = n(n+1)/2 = 6*7/2 = 21

```
Processing ReadCovarianceArray_new.C...
Event 0, number of tracks: 8
 Track 0 covariance:
   cov[0] = 0.0104456
   cov[1] = 2.366e-06
   cov[2] = 0.0103324
    cov[3] = -0.000289634
   cov[4] = 7.52669e-07
   cov[5] = 8.04972e-06
   cov[6] = -8.12589e-07
    cov[7] = 0.000284166
   cov[8] = 4.50984e-08
   cov[9] = 7.82997e-06
    cov[10] = 0.000153944
    cov[11] = 7.02629e-07
    cov[12] = -4.94218e-06
   cov[13] = 5.14138e-09
    cov[14] = 0.00011838
   cov[15] = -1.41347e-06
    cov[16] = -3.20263e-06
   cov[17] = 3.89044e-08
    cov[18] = -8.82041e-08
   cov[19] = 2.242e-09
```

```
Cov[0] = cov(10, 10)
                                    Cov[11] = cov(11, q/p)
                                                                                                              \sigma_{l1} = \sqrt{(Cov[2])}
                                                                                  \sigma_{l0} = \sqrt{(Cov[0])}
Cov[1] = cov(10, 11)
                                    Cov[12] = cov(\varphi, q/p)
Cov[2] = cov(I1, I1)
                                    Cov[13] = cov(\theta, q/p)
Cov[3] = cov(10, \varphi)
                                    Cov[14] = cov(q/p, q/p)
                                                                                  \sigma_{\phi} = \sqrt{(Cov[5])}
                                                                                                             \sigma_{\theta} = \sqrt{(Cov[9])}
cov[4] = cov(11, \varphi)
                                    Cov[15] = cov(10, time)
cov[5] = cov(\varphi, \varphi)
                                    Cov[16] = cov(11, time)
                                                                                  \sigma_q / p = \sqrt{(Cov[14])}
cov[6] = cov(10, \theta)
                                    Cov[17] = cov(\varphi, time)
cov[7] = cov(11, \theta)
                                    Cov[18] = cov(\theta, time)
cov[8] = cov(\varphi, \theta)
                                    Cov[19] = cov(q/p, time)
cov[9] = cov(\theta, \theta)
                                    Cov[20] = cov(time, time)
cov[10] = cov(10, q/p)
```

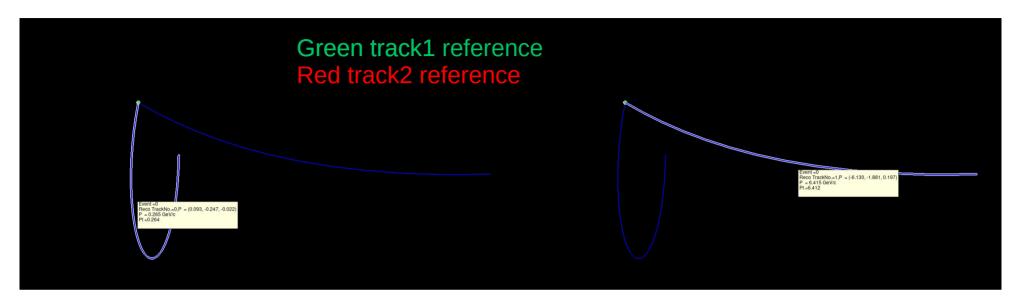
cov[20] = 0.000333566

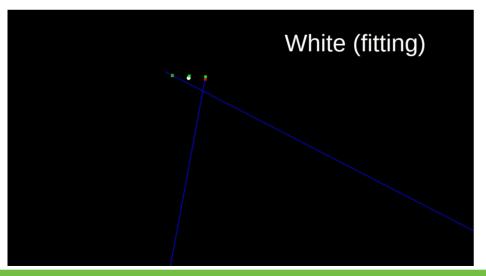
Event display (First two tracks)





Event display (First two tracks)





- All three methods (track 1 reference, track2 reference, and distance minimization) are compatible
- Minor difference is due to analytical approach
- Distance minimization returns the unique point