

# Updates on $D^0$ Reconstruction and Tracking Performance

**Shyam Kumar\***, Annalisa Mastroserio, Domenico Elia  
INFN Bari, Italy

# Secondary Vertex Reconstruction

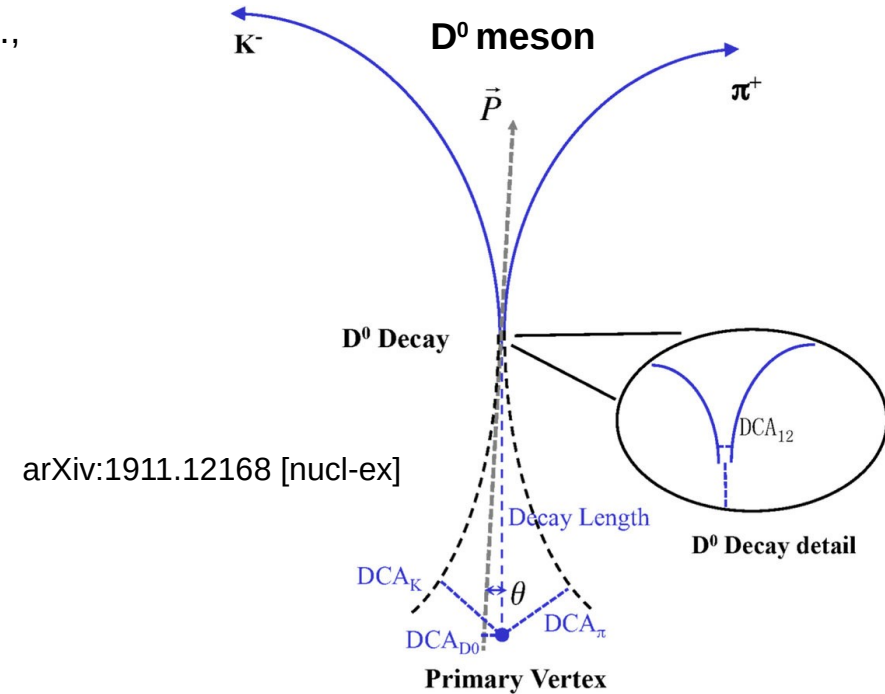
**Secondary vertex:** allows access to more topological variables, e.g., decay length ( $dl$ ), pointing angle ( $\cos\theta$ ),  $DCA_{D^0}$  or  $DCA_{\Lambda_c}$  etc.

## Topological Variables

$$\vec{dl} = \vec{S}\vec{V} - \vec{P}\vec{V}$$

$$\cos\theta = \frac{\vec{dl} \cdot \vec{p}_{D^0}}{|\vec{dl}| |\vec{p}_{D^0}|}$$

$$DCA_{D^0} = |\vec{dl}| \sin\theta$$



## Secondary Vertex Reconstruction:

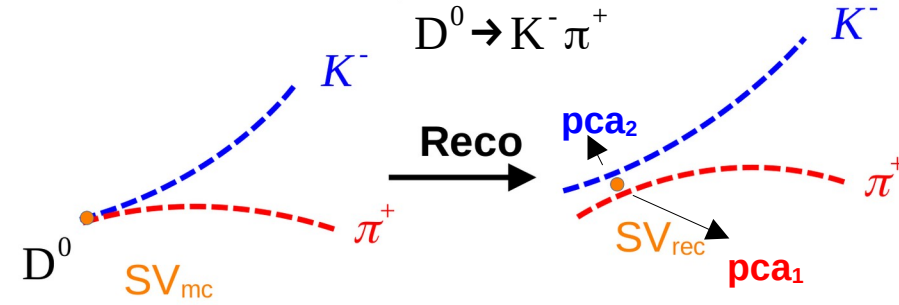
- **Helix Swimming (Ignores track errors)**
- **Chi2 minimization (considers track errors): Shyam**
- **AdaptiveMultiVertexFinder (considers track errors): In development: Bishoy**
- **KFParticle (considers track errors): In development: Xin and others**

Based on  
Kalman filter

# Secondary Vertex Resolution

$$\vec{SV} = \frac{(p\vec{ca}_1 + p\vec{ca}_1)}{2} \quad SV_x = \frac{(pca_{1x} + pca_{1x})}{2}$$

$$\sigma_{SV_x}^2 = \frac{(\sigma_{pca_{1x}}^2 + \sigma_{pca_{2x}}^2 + 2 \text{cov}(pca_{1x}, pca_{2x}))}{4}$$



Fit Range: Mean +/- 2\*StdDev

10.2025

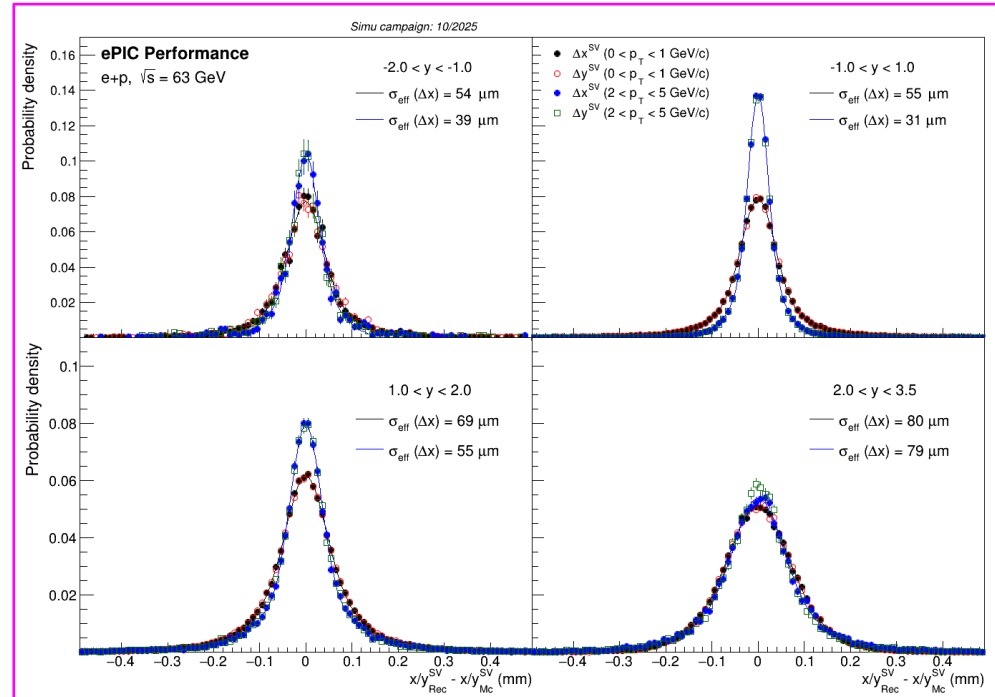
preTDR

Chi-square minimization

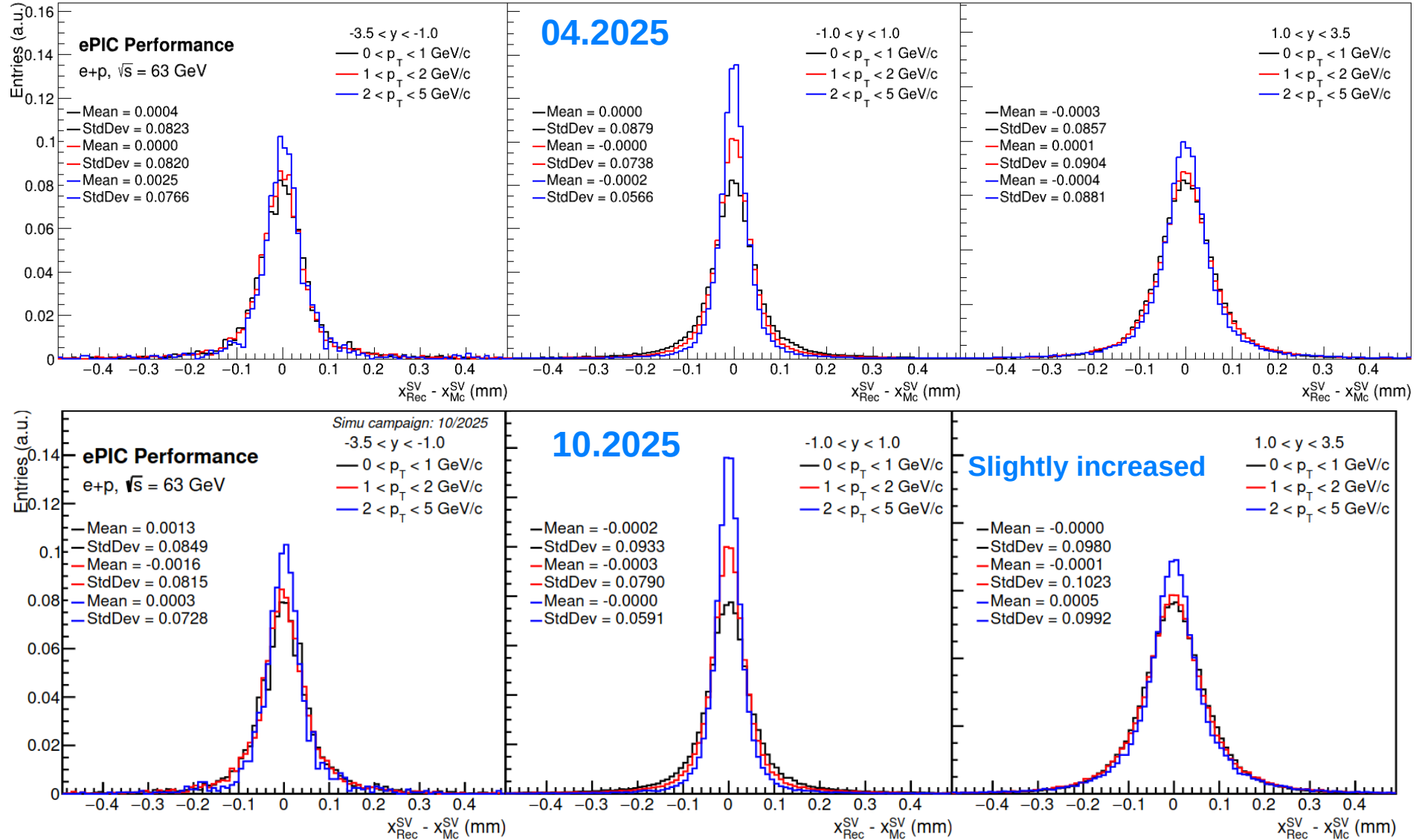
Two Gaussian ( $A_{\text{core}}$ ,  $\sigma_{\text{core}}$ ,  $A_{\text{tail}}$ ,  $\sigma_{\text{tail}}$ )

$$\sigma_{\text{eff}} = \sqrt{\frac{A_{\text{core}}}{(A_{\text{core}} + A_{\text{tail}})} \sigma_{\text{core}}^2 + \frac{A_{\text{tail}}}{(A_{\text{core}} + A_{\text{tail}})} \sigma_{\text{tail}}^2}$$

$\sigma_{\text{eff}}$  decreases as we go at high  $p_T$

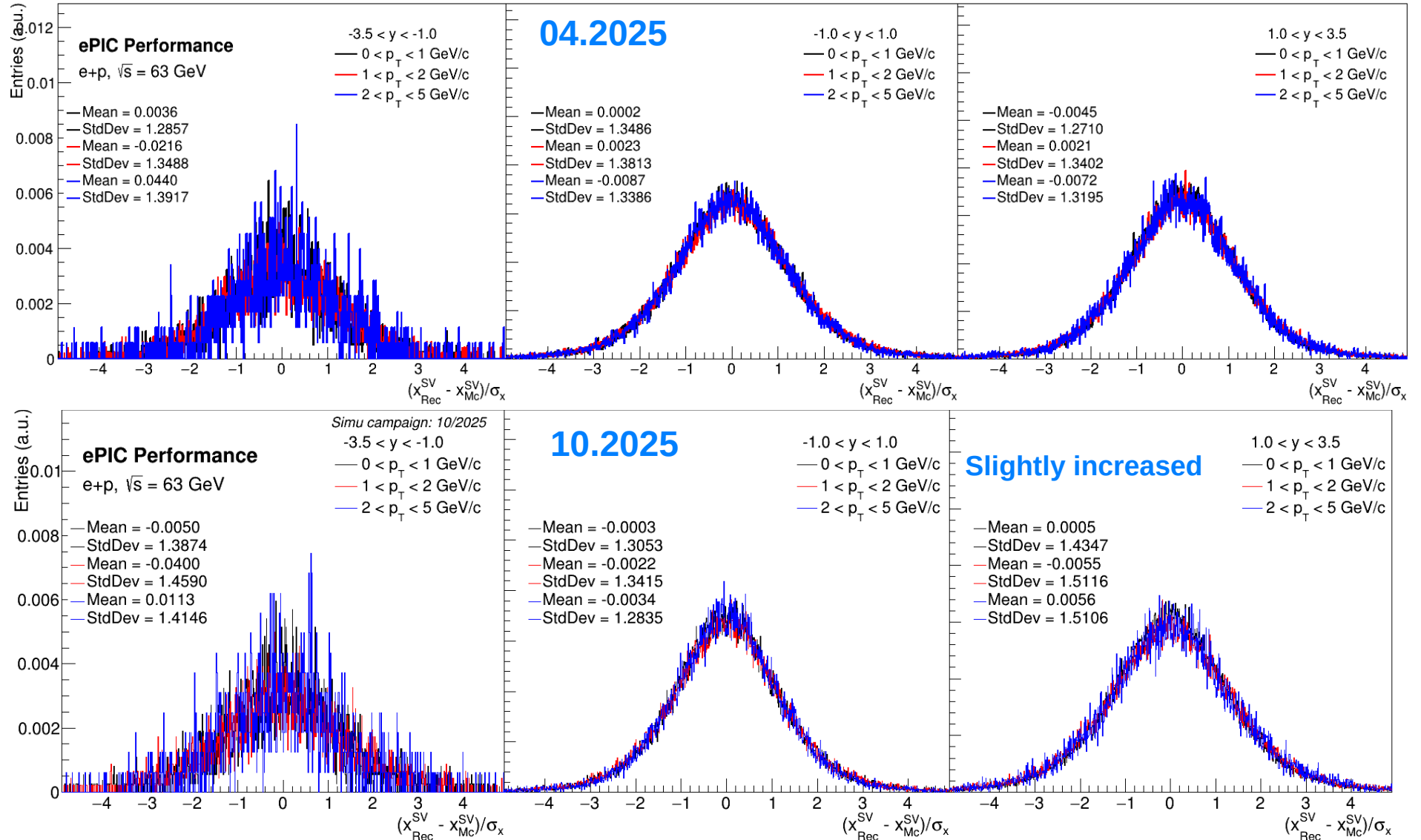


# Secondary Vertex Resolution





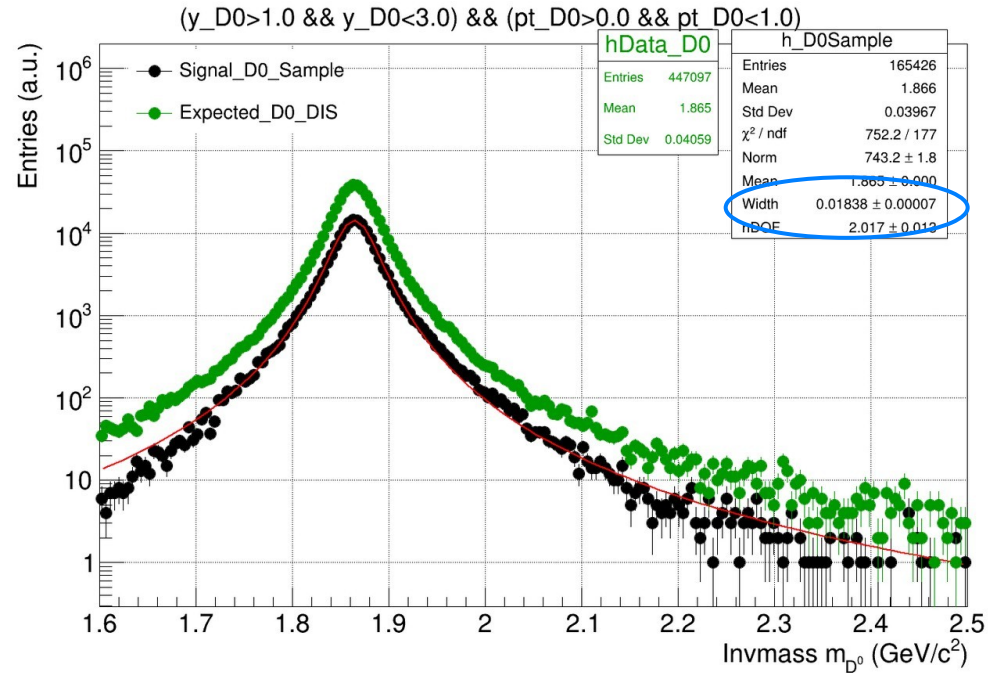
# Secondary Vertex Resolution



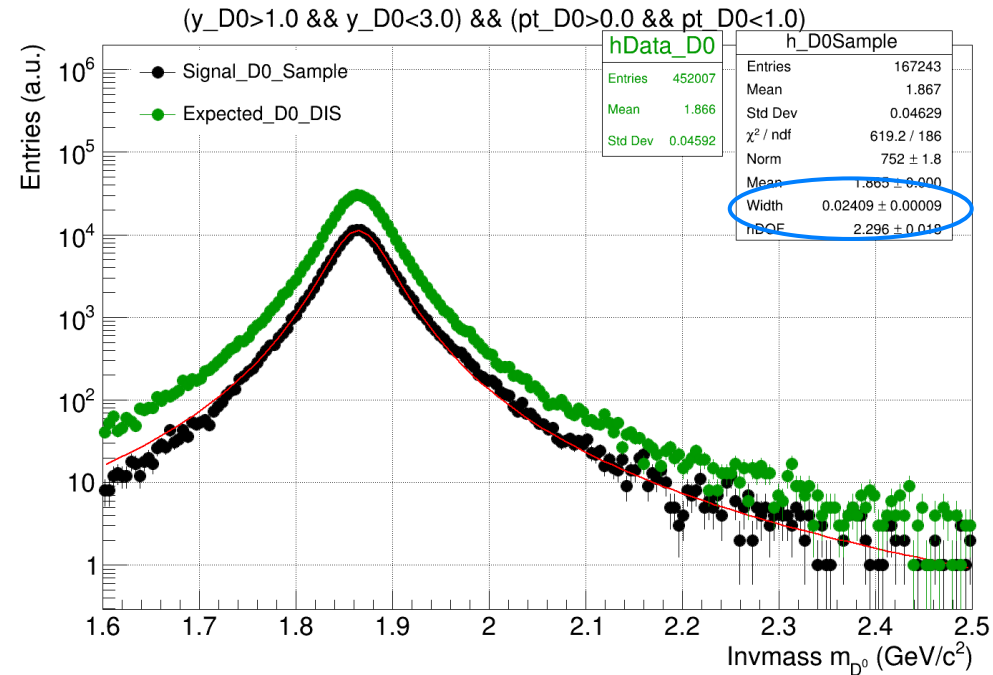
# Invariant mass

ep,10 x 100,  $Q^2 > 1 \text{ GeV}^2$

25.04.0



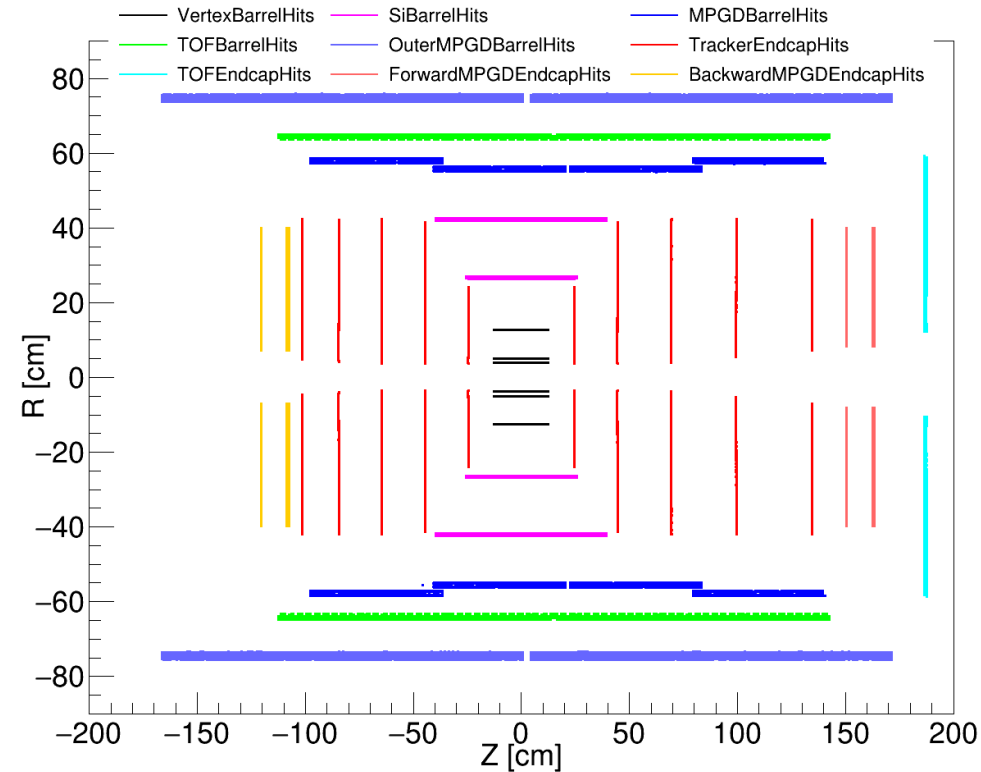
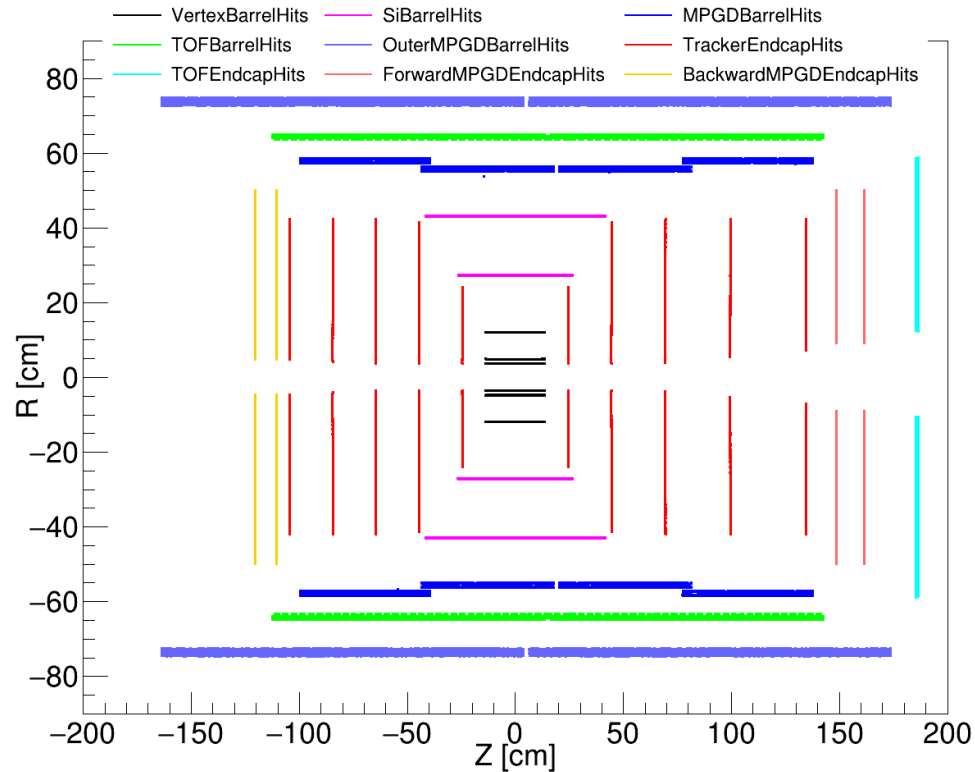
25.10.3



Black markers are from  $D^0$  Sample (Green sampled to  $10 \text{ fb}^{-1}$ )

# Tracking Geometry

## Single particle (pi-) simulation

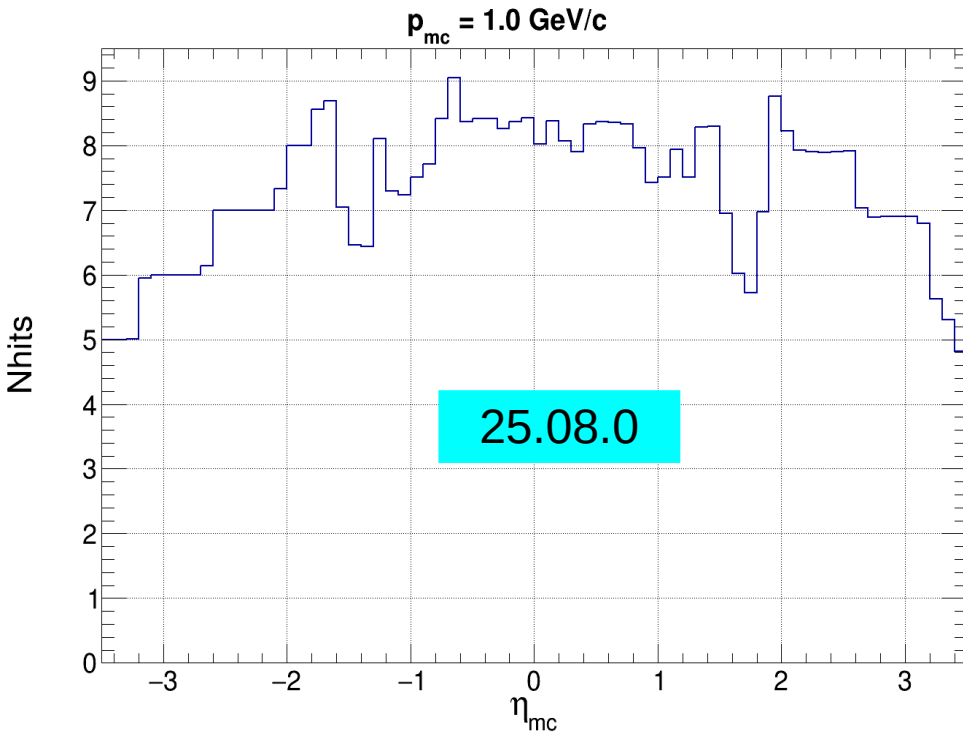


Forward/backward MPGDs have been modified, including changes in their thickness

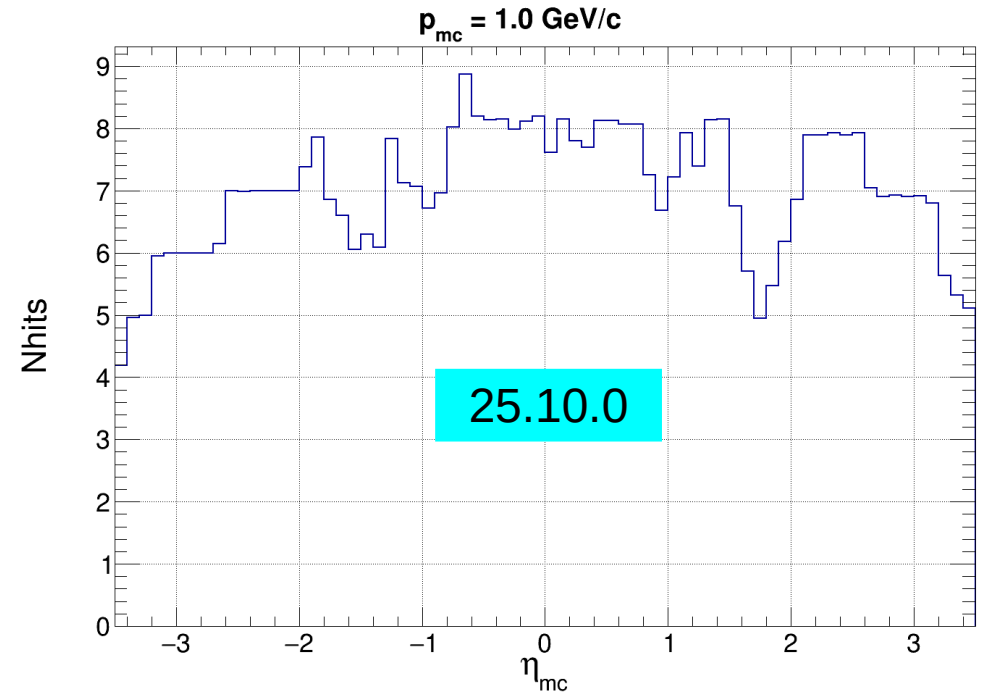
The thickness of SVT outer barrel (OB) layers has also increased

## Single particle ( $\pi^-$ ) simulation

Nhits at MC Level (no digitization)



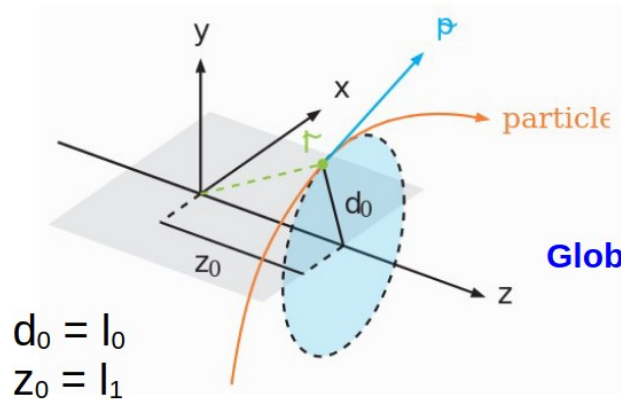
Nhits at MC Level (no digitization)



# Track Parameters in ACTS

[https://indico.bnl.gov/event/28544/contributions/109057/attachments/62799/108633/ePIC\\_Tracking\\_Meeting\\_30June2025\\_ShyaamKumar.pdf](https://indico.bnl.gov/event/28544/contributions/109057/attachments/62799/108633/ePIC_Tracking_Meeting_30June2025_ShyaamKumar.pdf)

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



At Point of closest approach

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

Global (Lab frame) ↓

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

$$x = -l_0 \sin \phi, \quad y = l_0 \cos \phi, \quad z = l_1$$

$$p_x = p \cos \phi \sin \theta, \quad p_y = p \sin \phi \sin \theta, \quad p_z = p \cos \theta$$

$$\text{charge} = \text{sign}(q/p)$$

Plan to add it to the benchmarks

$$\text{Pull } l_0 = \frac{(l_{0\text{rec}} - l_{0\text{gen}})}{\sigma_{l_0}}$$

$$\text{Pull } \phi = \frac{(\phi_{\text{rec}} - \phi_{\text{gen}})}{\sigma_{\phi}}$$

$$\text{Pull } q/p = \frac{(q/p_{\text{rec}} - q/p_{\text{gen}})}{\sigma_{q/p}}$$

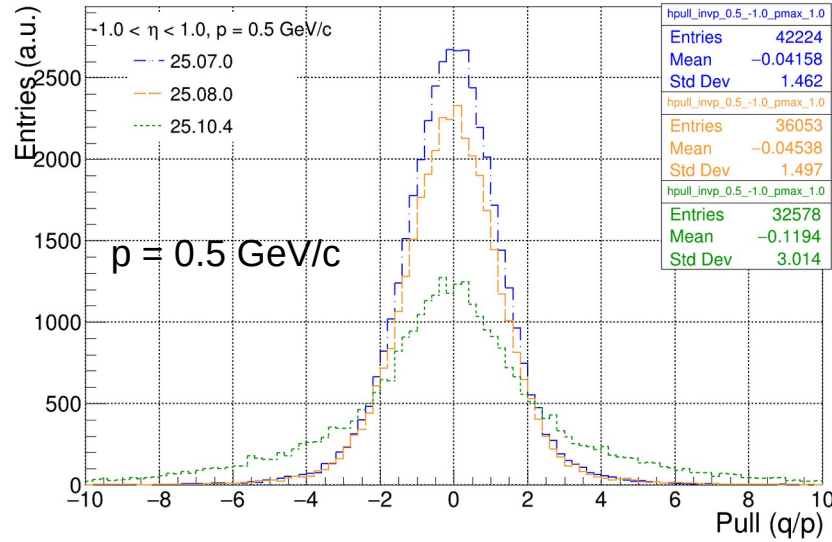
$$\text{Pull } l_1 = \frac{(l_{1\text{rec}} - l_{1\text{gen}})}{\sigma_{l_1}}$$

$$\text{Pull } \theta = \frac{(\theta_{\text{rec}} - \theta_{\text{gen}})}{\sigma_{\theta}}$$

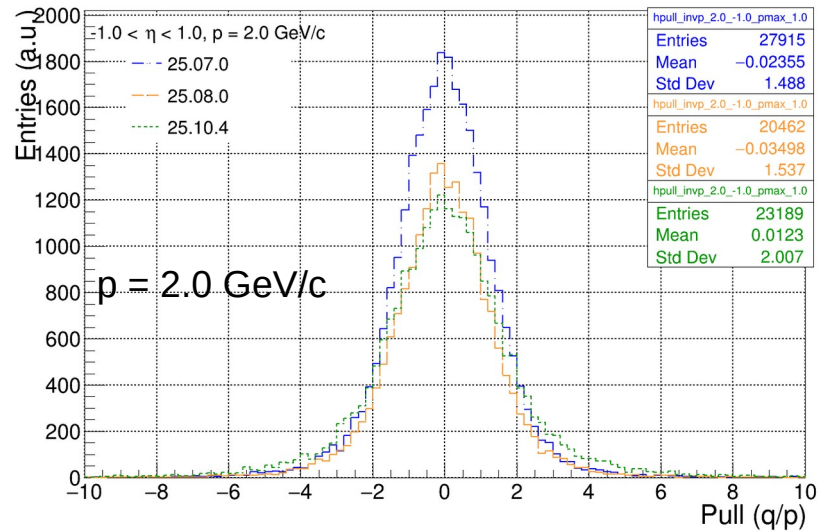
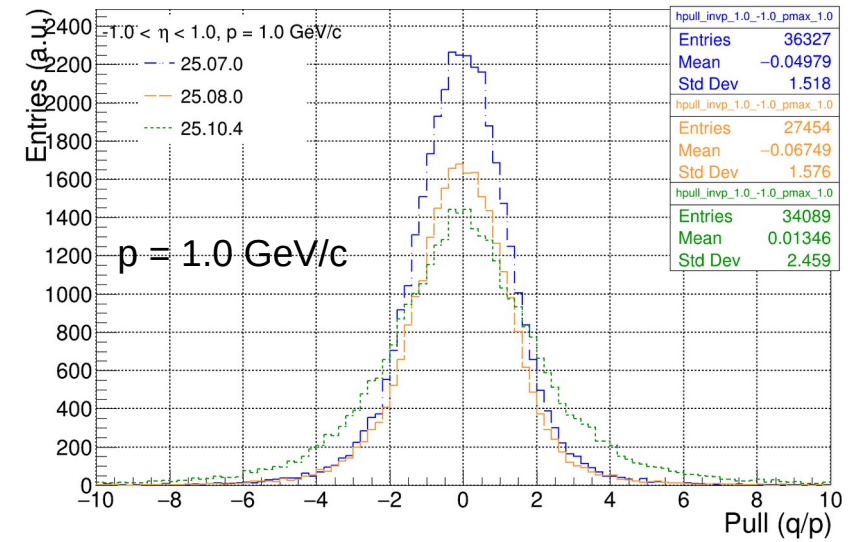
$$\text{Pull } p = \frac{(p_{\text{rec}} - p_{\text{gen}})}{\sigma_p}$$

`TTreeReaderArray<std::array<float, 21>> rcTrkCov(myReader, "CentralCKFTrackParameters.covariance.covariance[21]");` // Reading covariance this way works but gives an error message (otherwise change it to vector in data type)

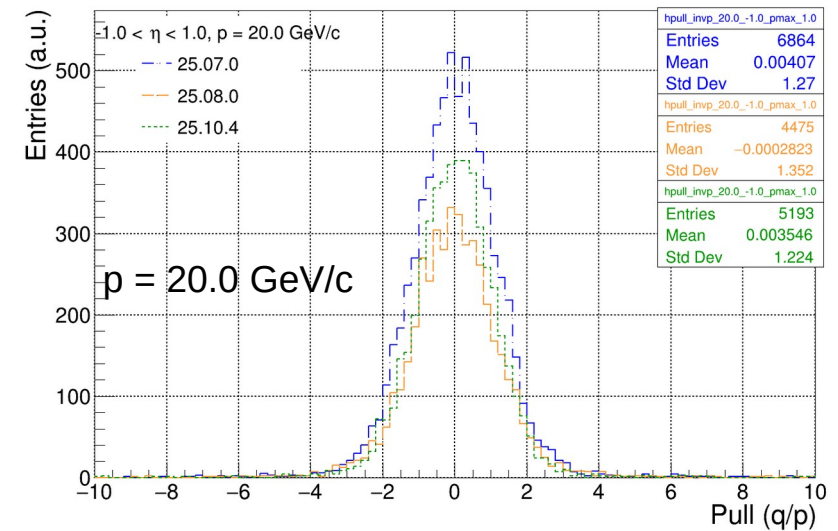
# Pull distributions (q/p)



$-1 < \eta < 1$

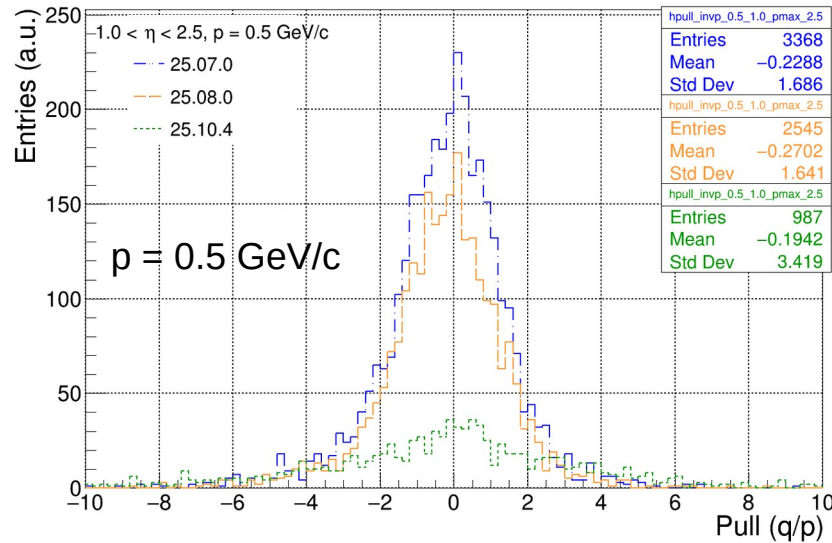


Tracking  
benchmarks

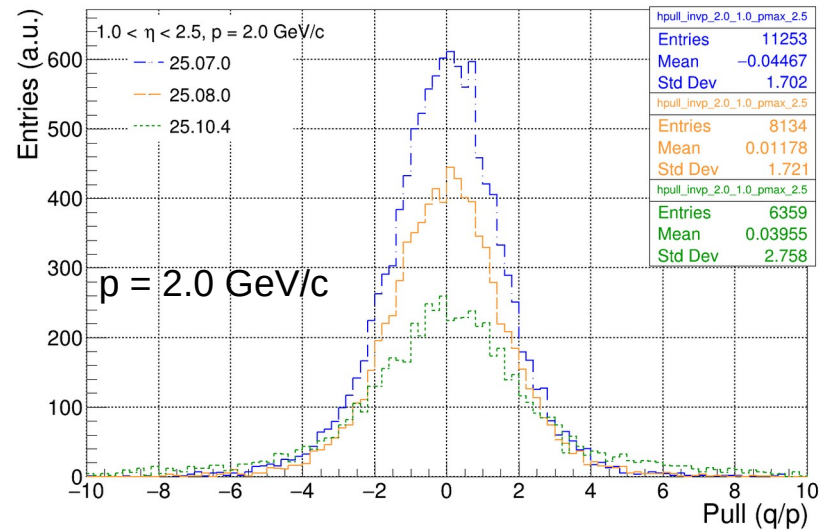
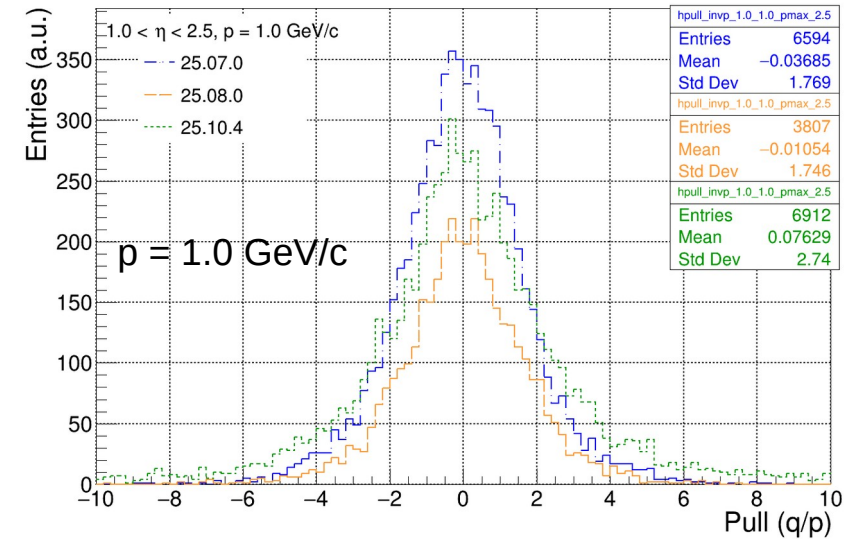




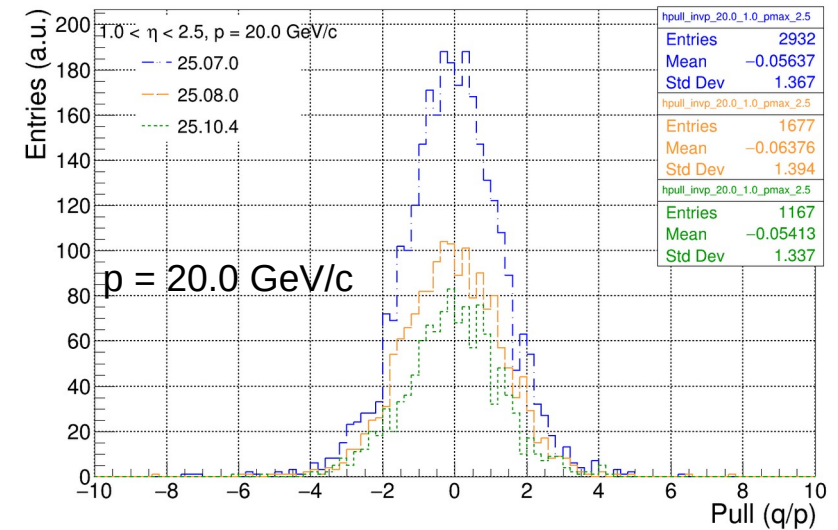
# Pull distributions (q/p)



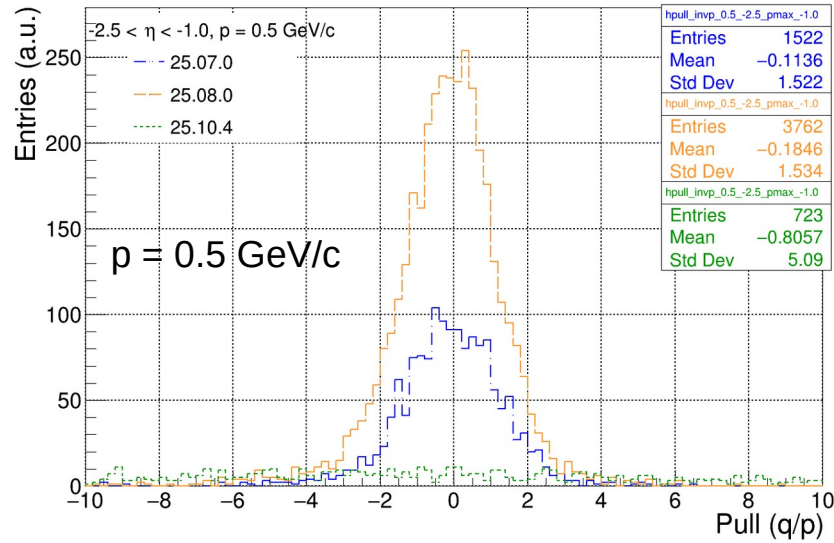
1.0 <  $\eta$  < 2.5



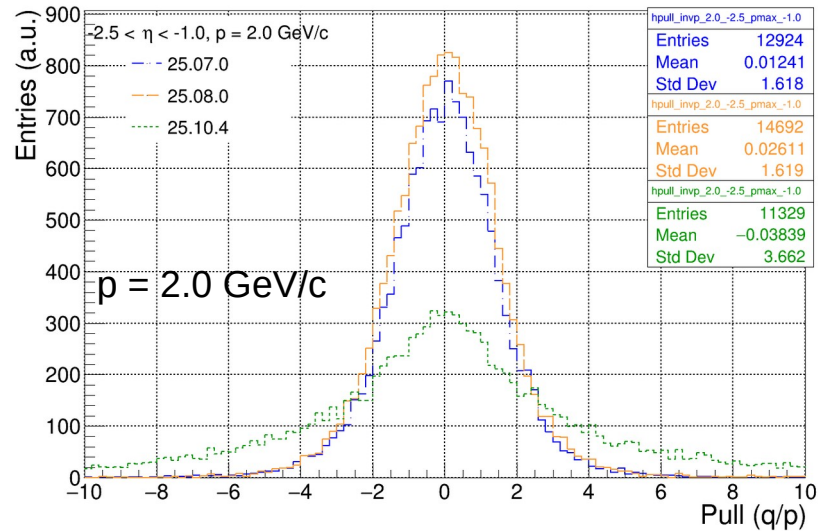
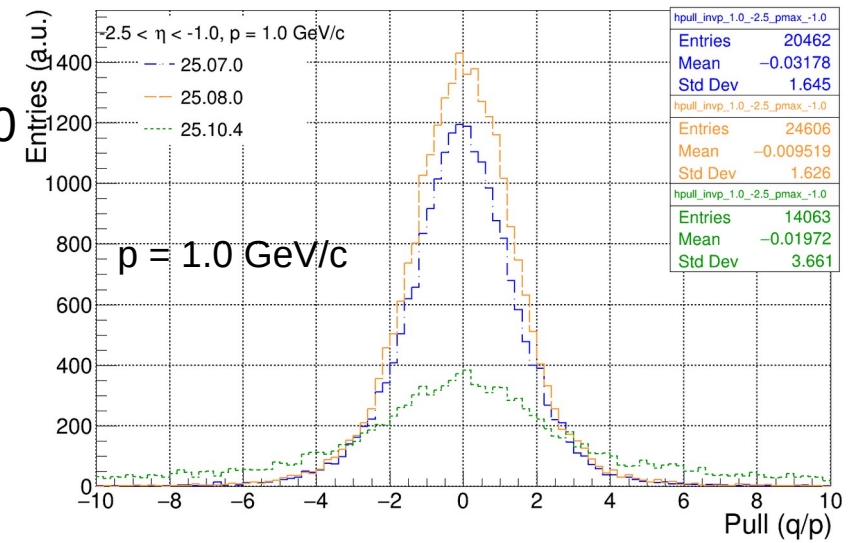
Tracking  
benchmarks



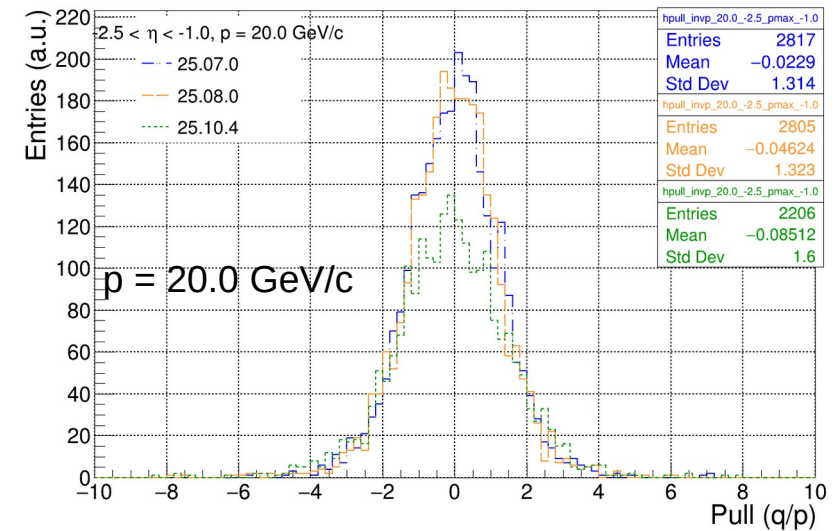
# Pull distributions (q/p)



$-2.5 < \eta < -1.0$



Tracking  
benchmarks



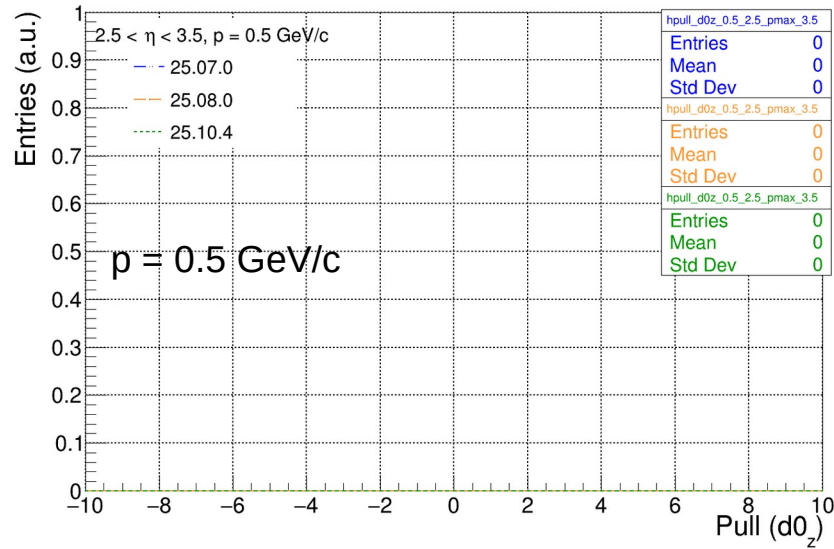


# Summary

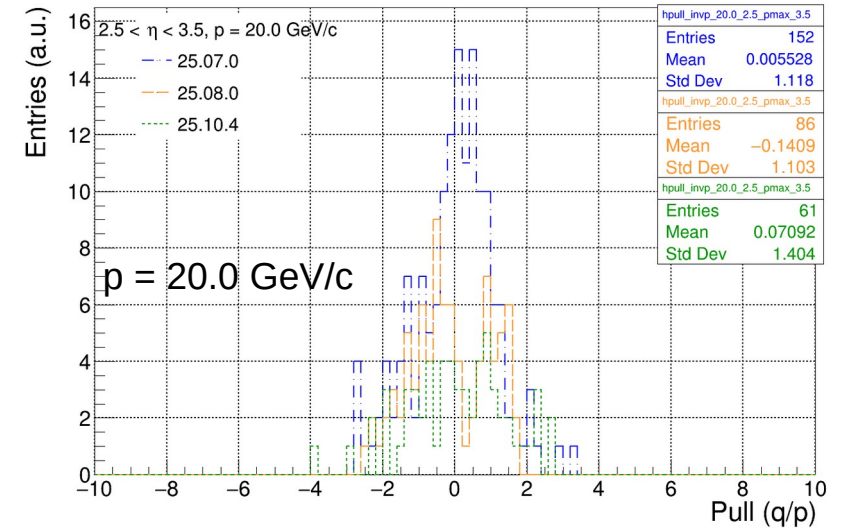
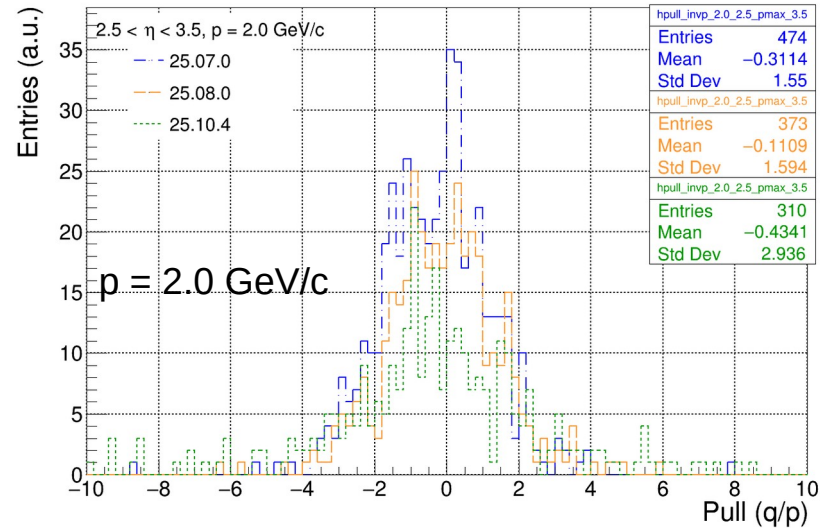
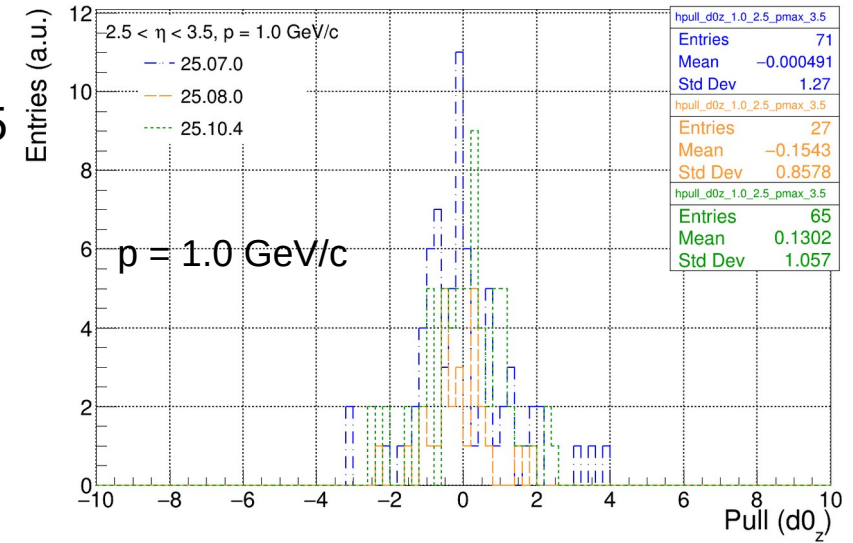
- There were many changes in the detector geometry during the October campaign.
- A small degradation was observed in the secondary-vertex resolution, which is expected since it depends on the  $DCA_{xy}$  resolution.
- The pulls for momentum are significantly degraded, indicating a mismatch between the material map and the updated geometry.
- Future Steps:
  - I have already added the momentum pulls to the benchmark tests, which must be checked at every commit.
  - The momentum pulls must remain consistent and close to  $\sim 1$ .
  - Before starting each campaign, produce a single-particle simulation and validate the pulls and related performance metrics.
  - It would be useful to present a comparison of these performances across consecutive campaigns.

**Thank you for your attention!**

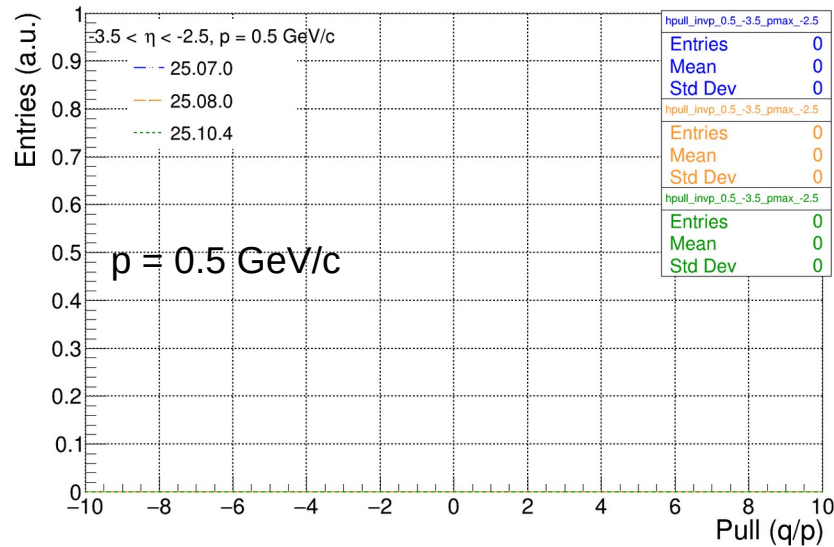
# Pull distributions (q/p)



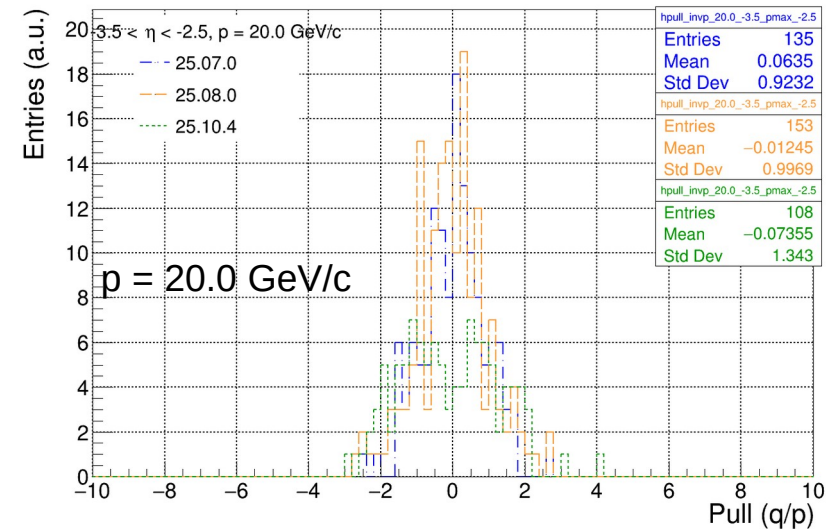
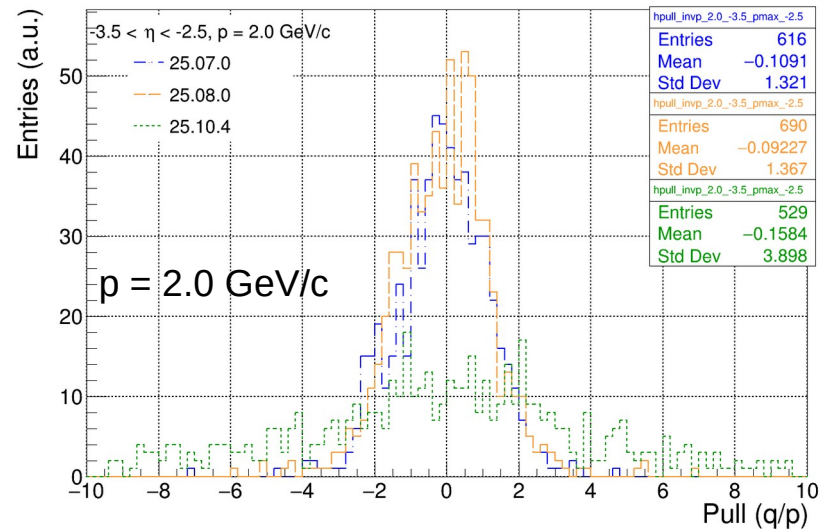
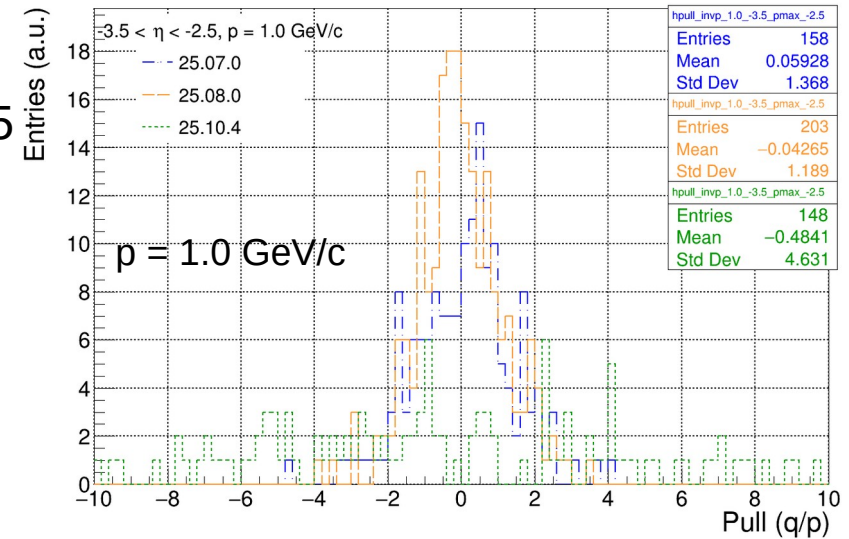
2.5 <  $\eta$  < 3.5



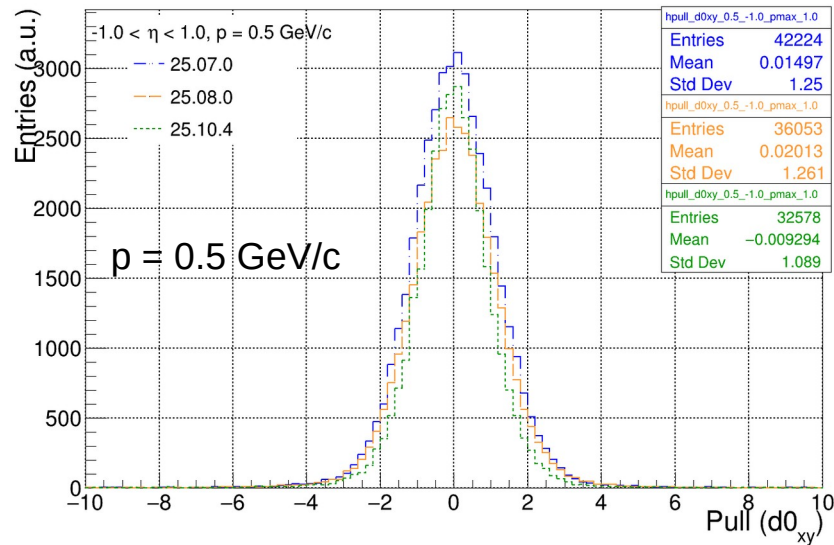
# Pull distributions (q/p)



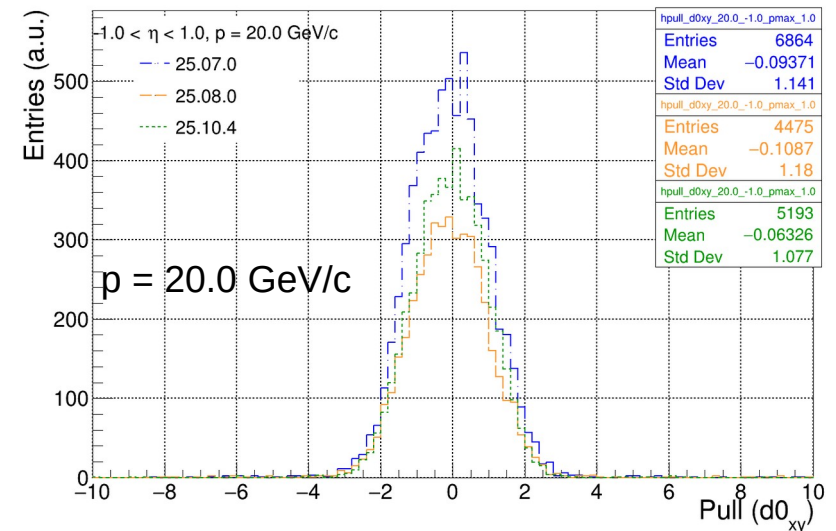
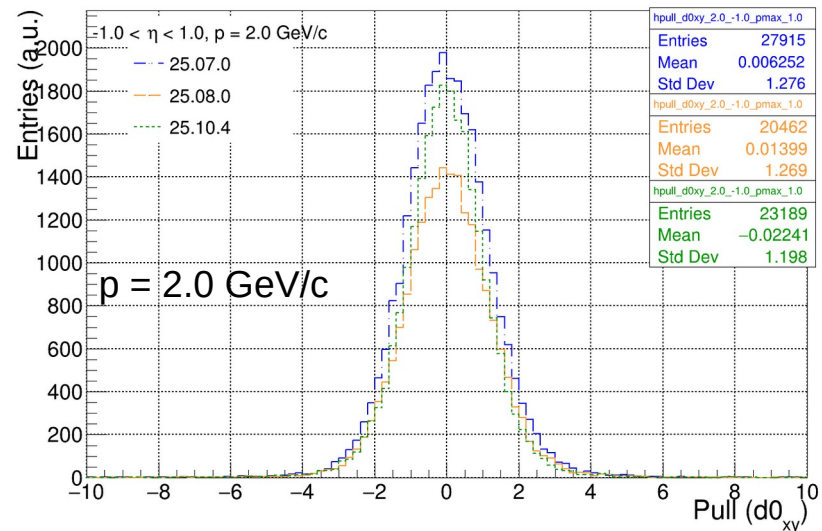
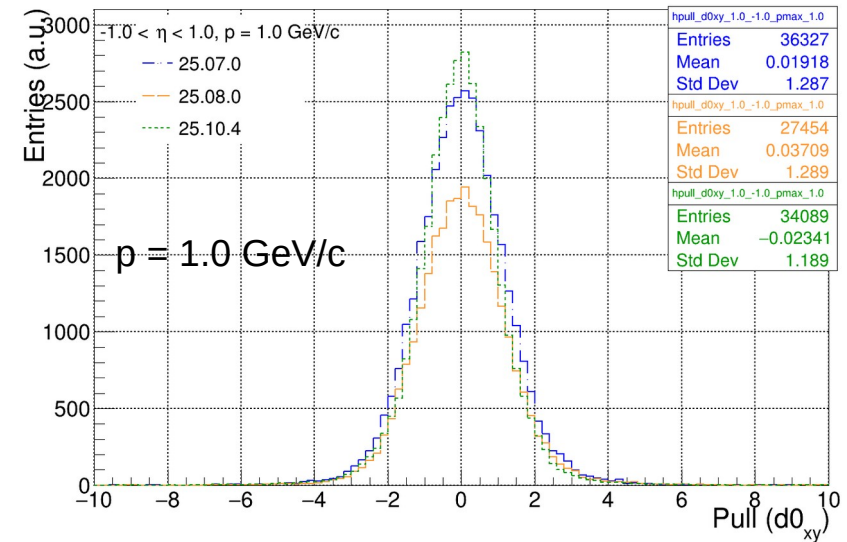
$-3.5 < \eta < -2.5$



# Pull distributions ( $d_{0xy}$ )



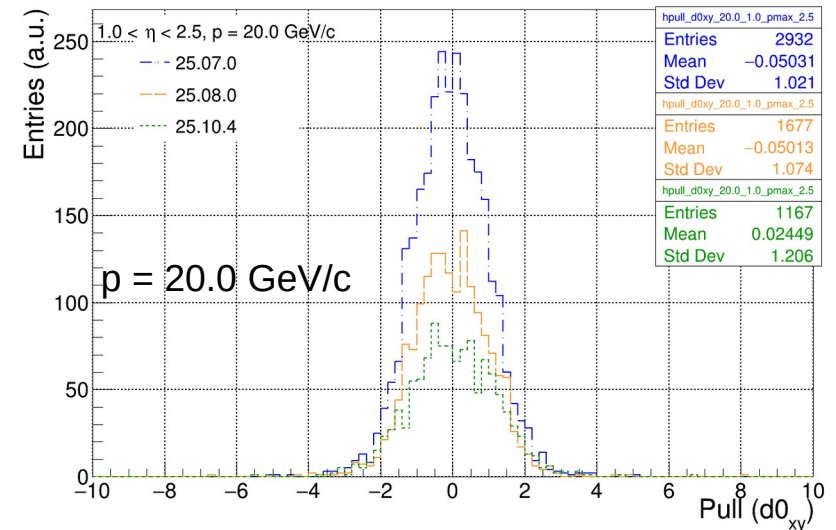
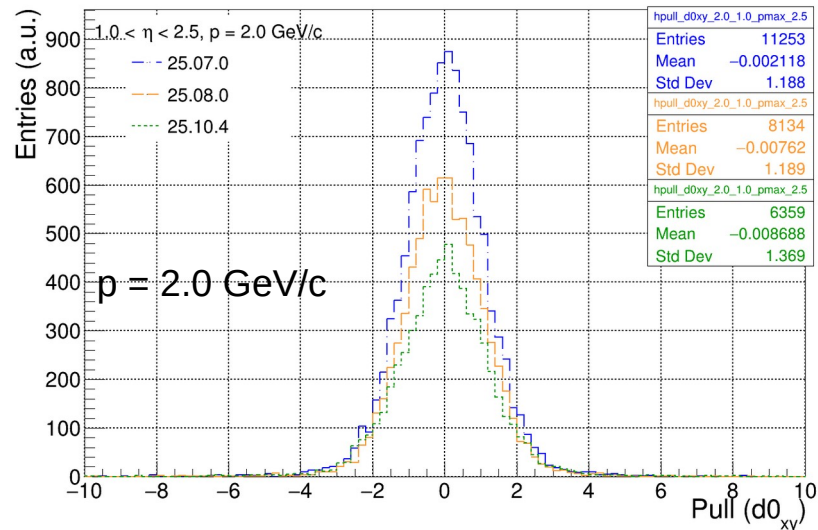
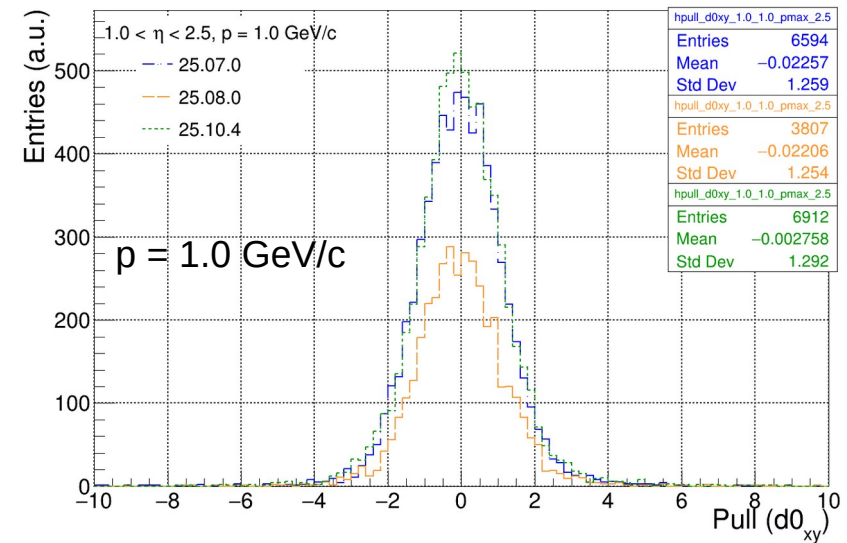
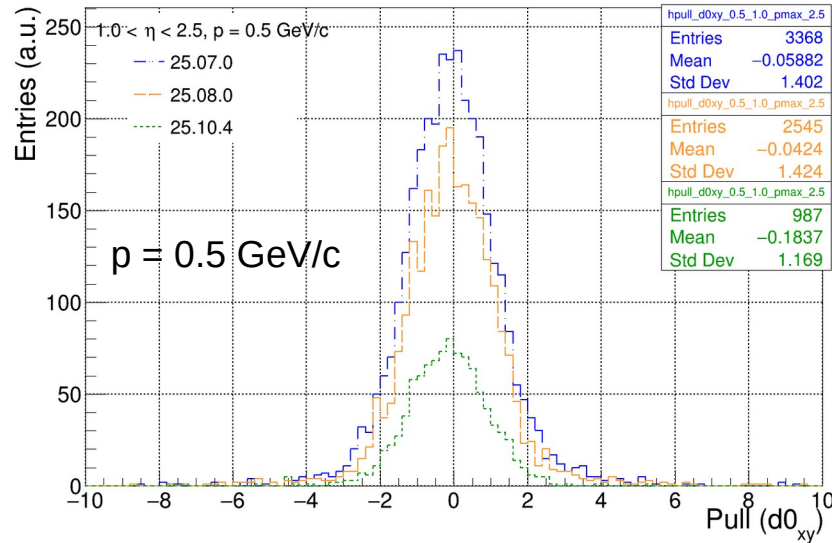
$-1 < \eta < 1$



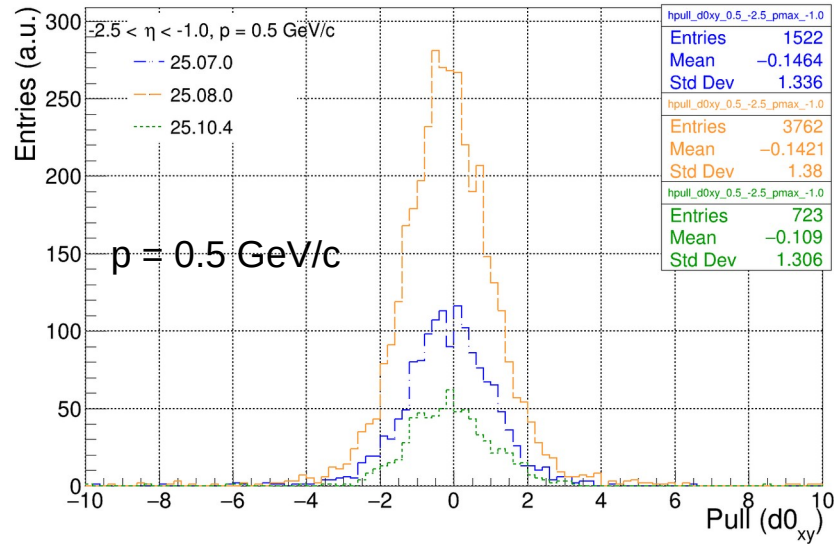


# Pull distributions ( $d_{0xy}$ )

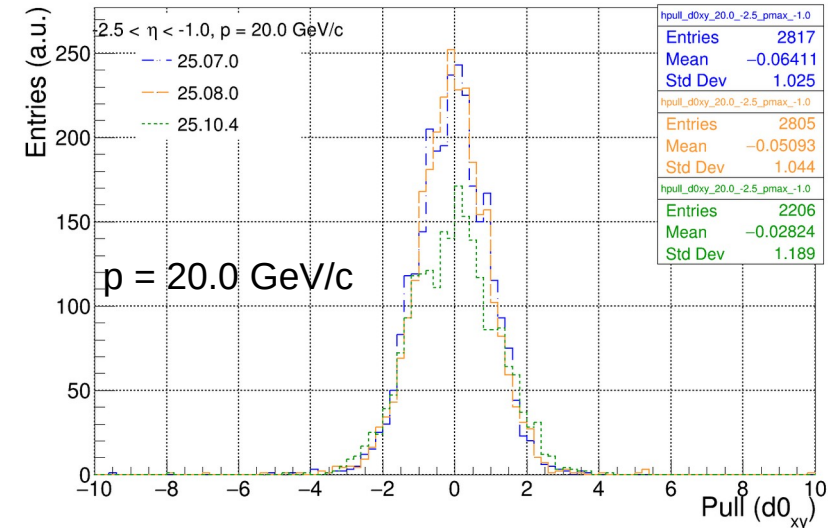
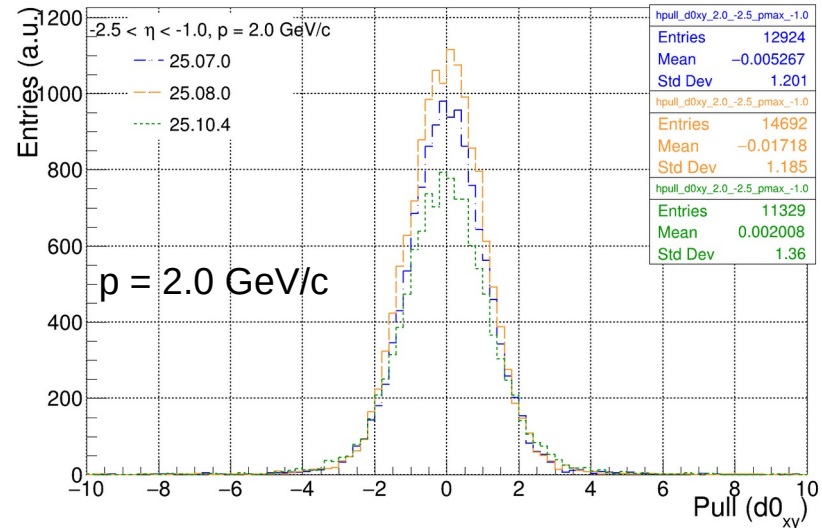
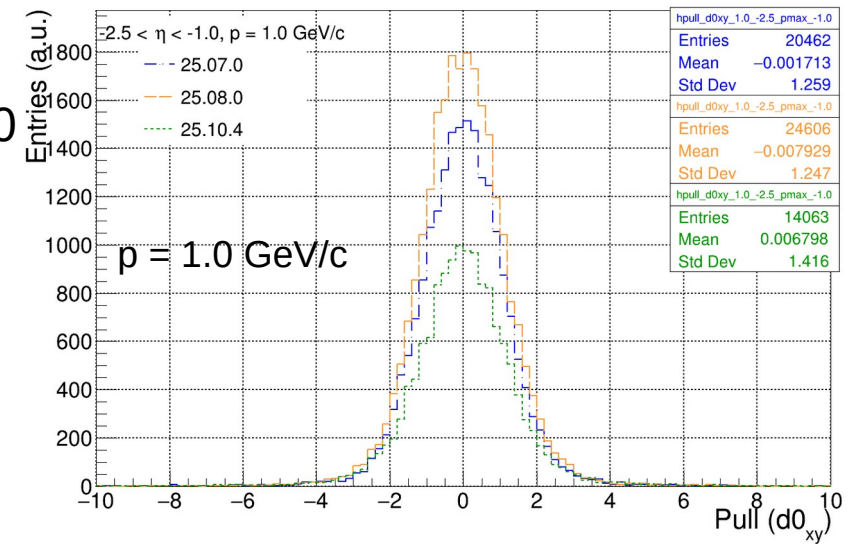
$1.0 < \eta < 2.5$



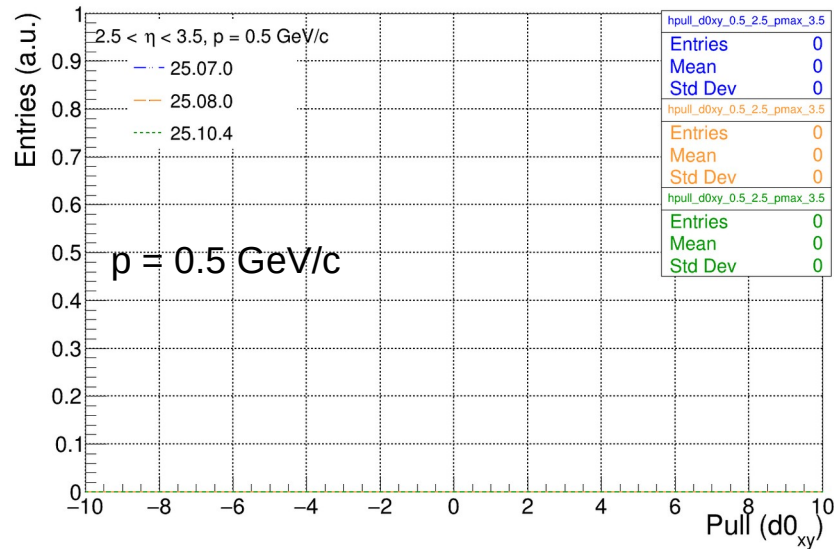
# Pull distributions ( $d_{0xy}$ )



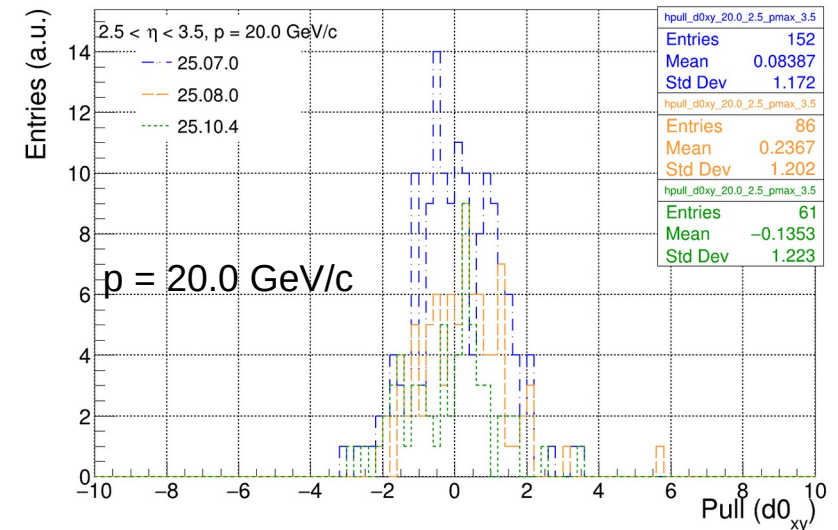
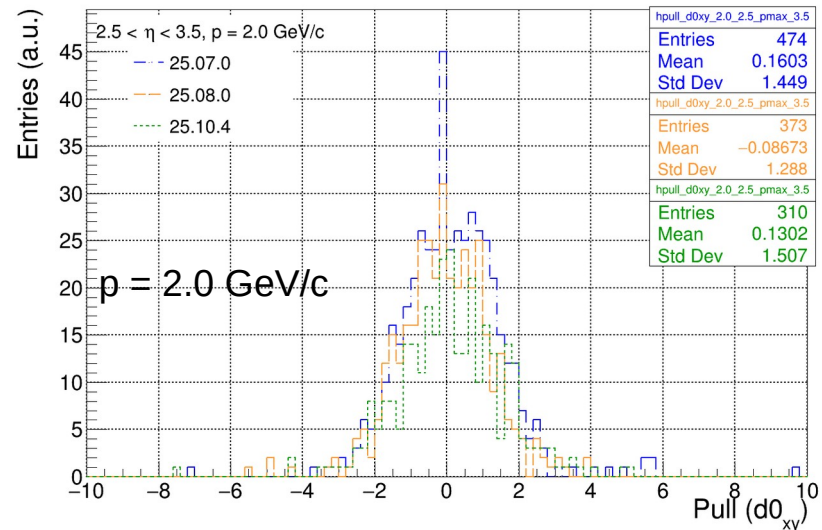
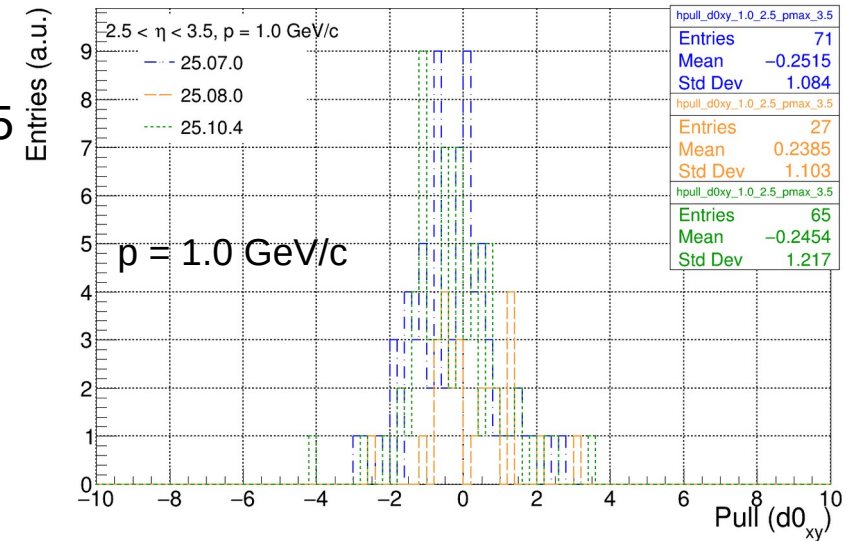
$-2.5 < \eta < -1.0$



# Pull distributions ( $d_{0xy}$ )

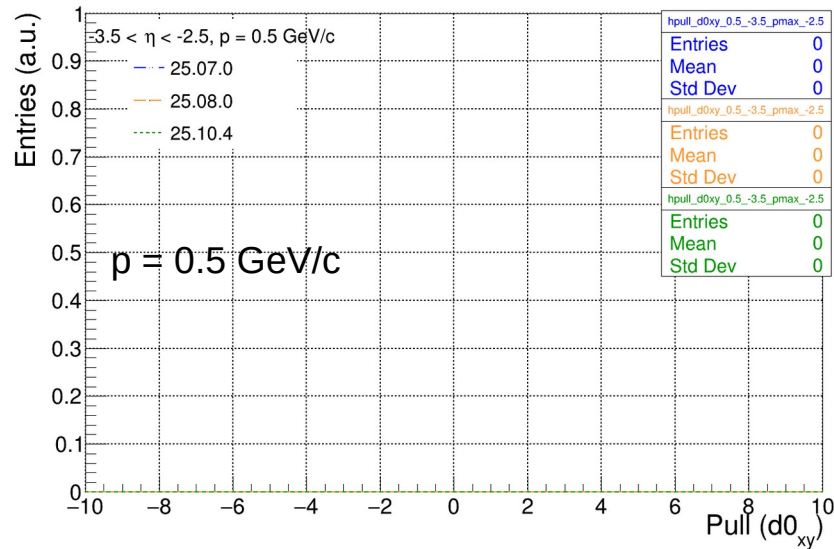


2.5 <  $\eta$  < 3.5

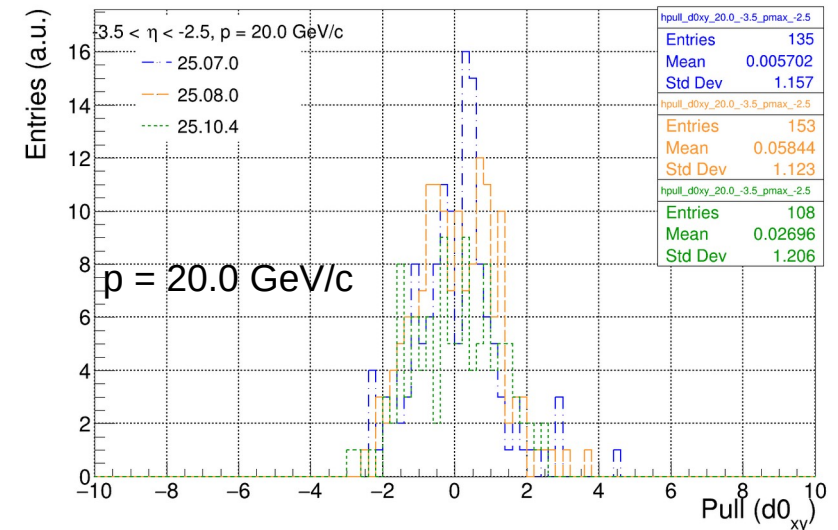
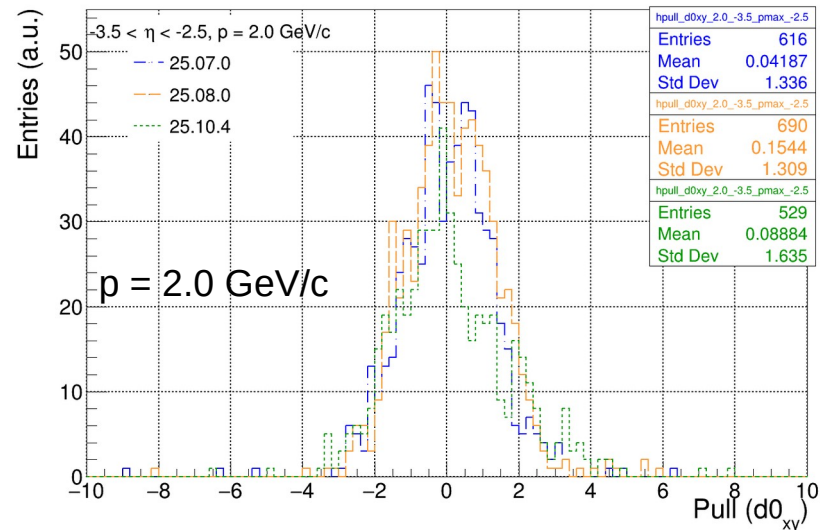
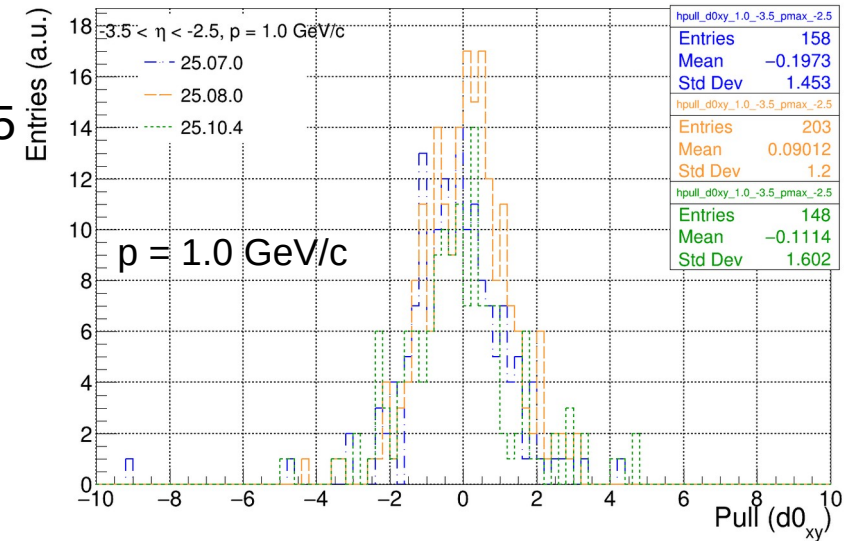




# Pull distributions ( $d_{0xy}$ )

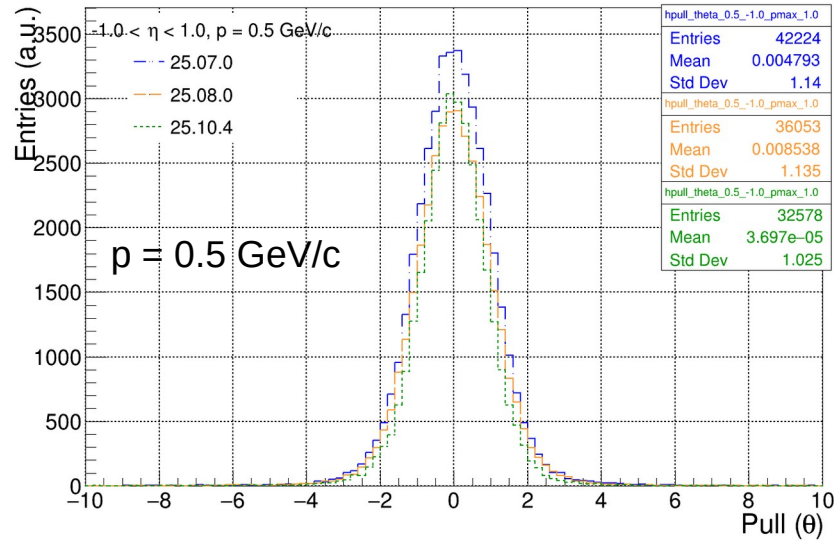


$-3.5 < \eta < -2.5$

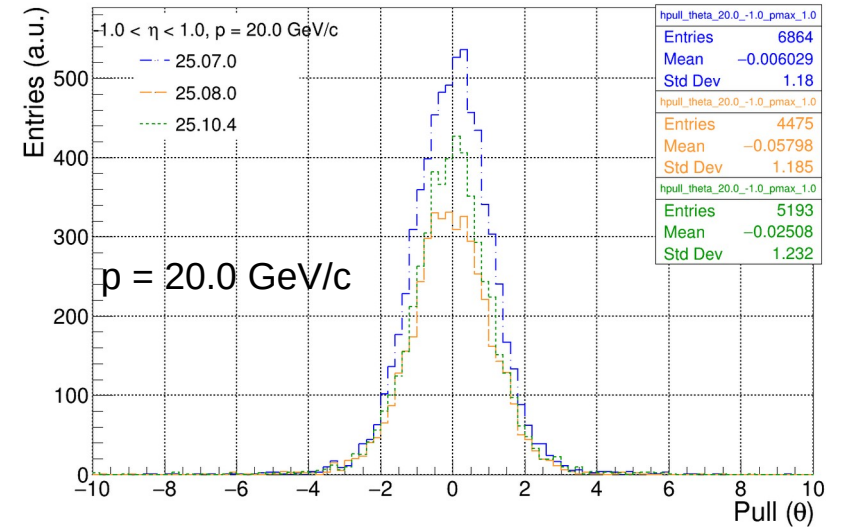
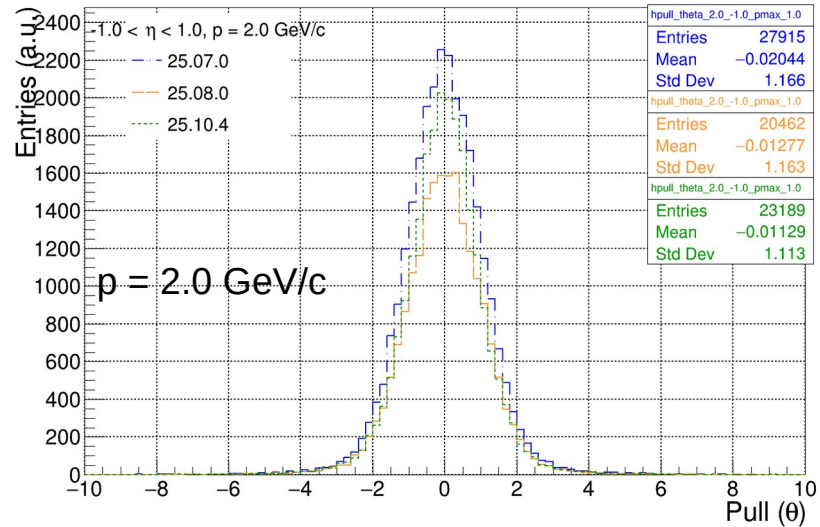
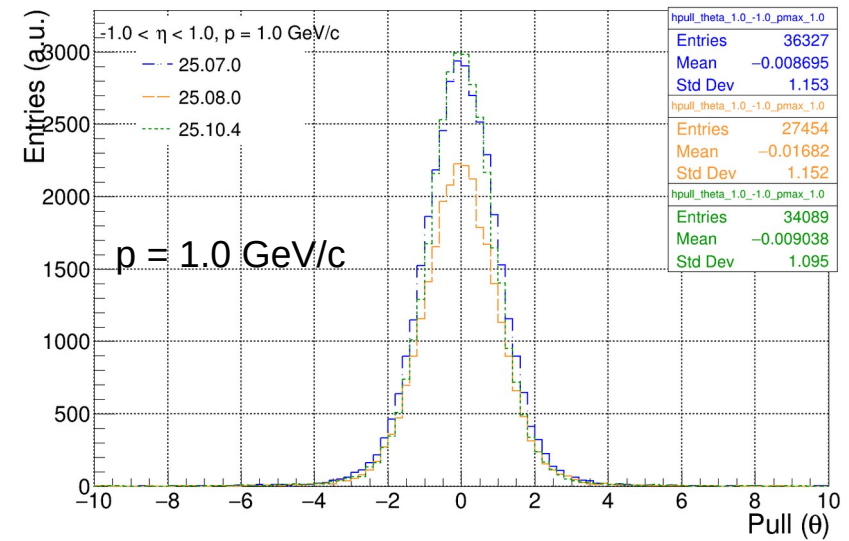




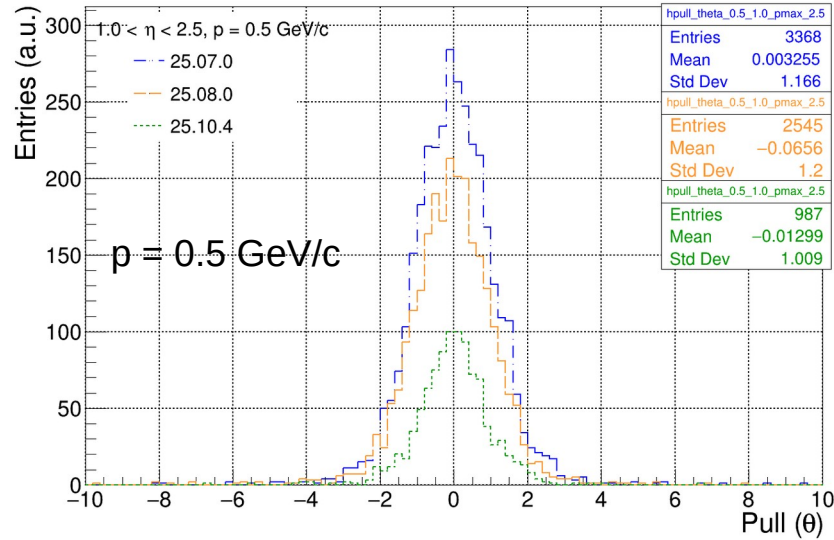
# Pull distributions ( $\theta$ )



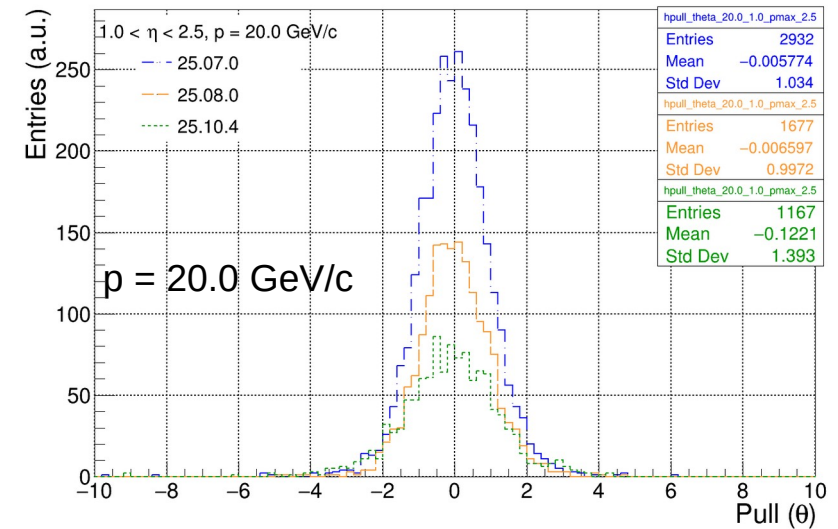
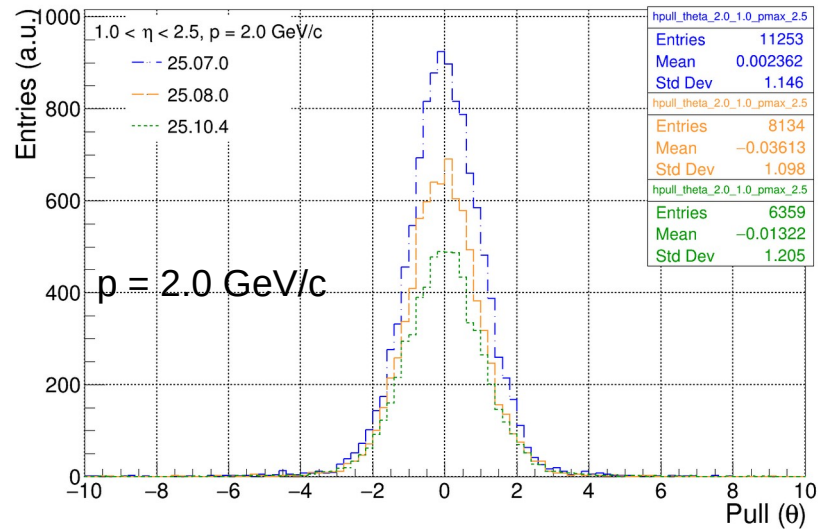
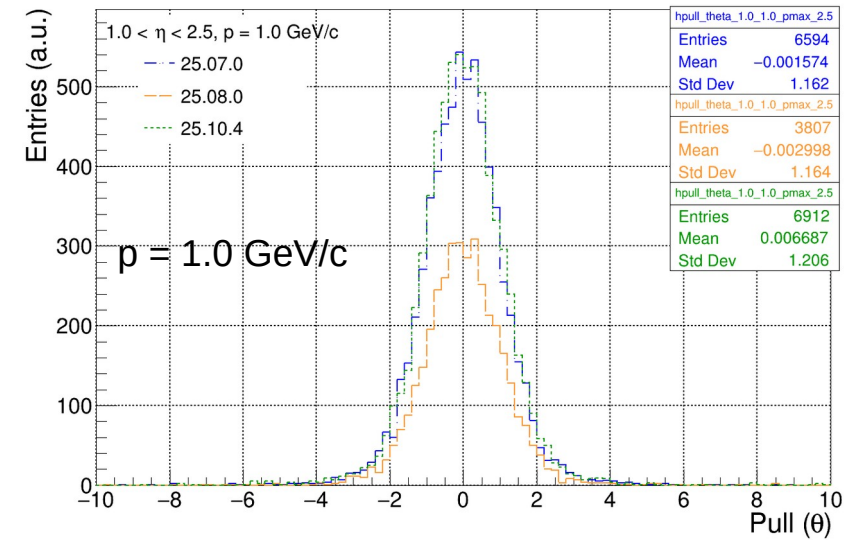
$-1 < \eta < 1$



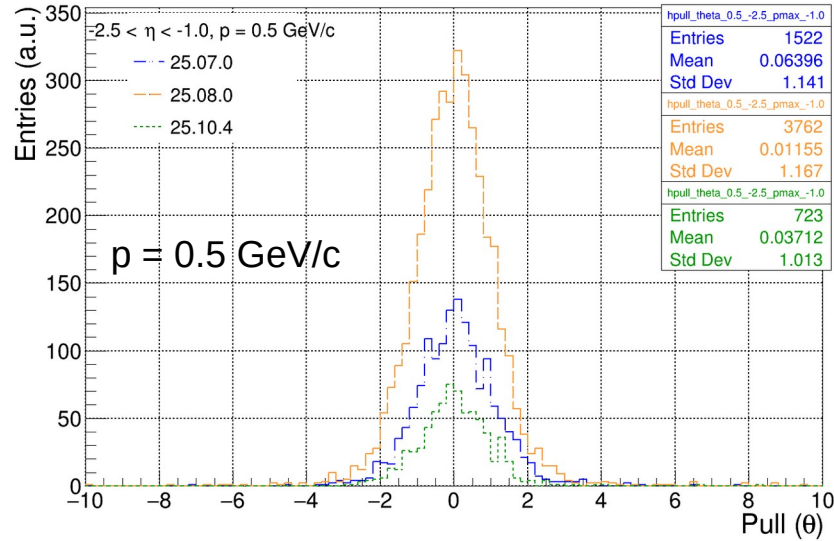
# Pull distributions ( $\theta$ )



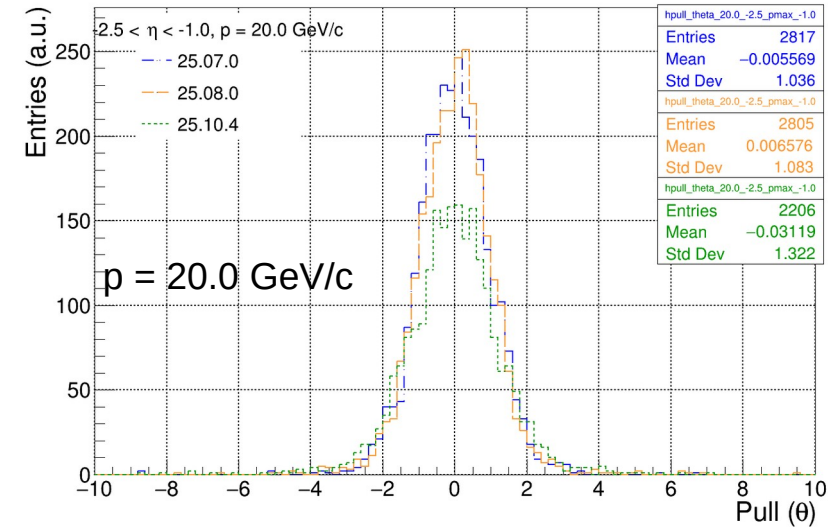
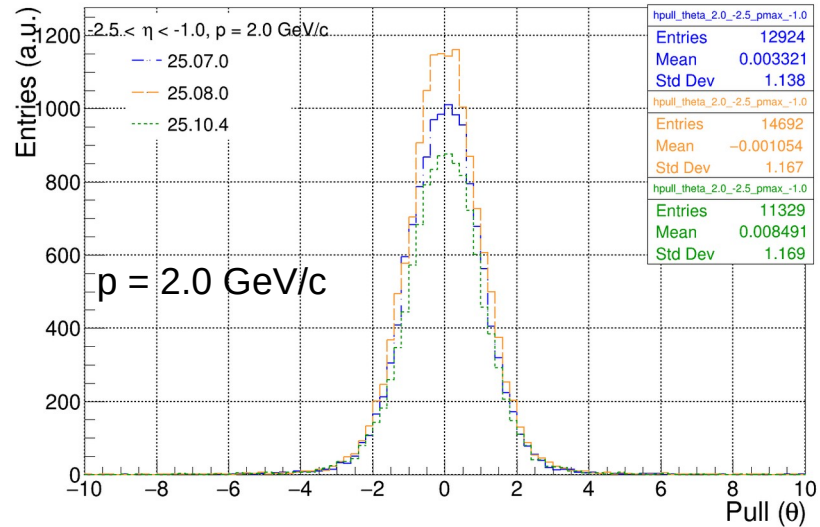
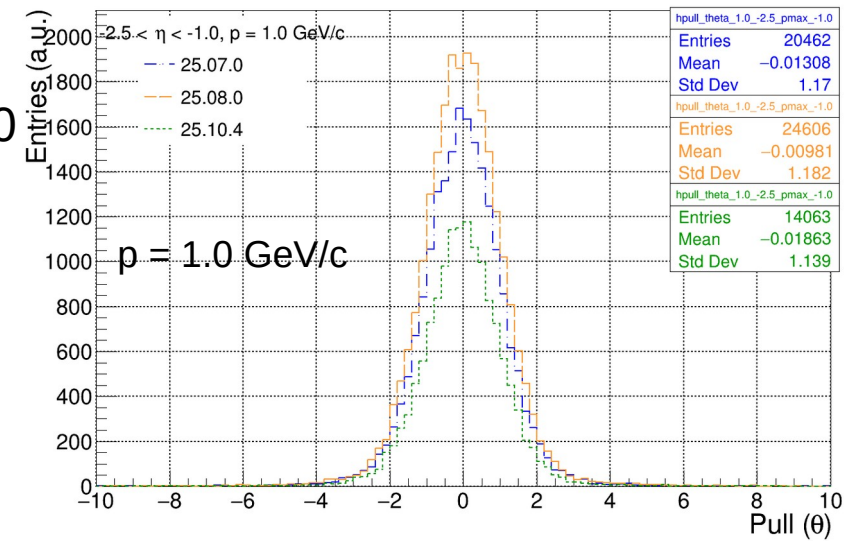
1.0 <  $\eta$  < 2.5



# Pull distributions ( $\theta$ )

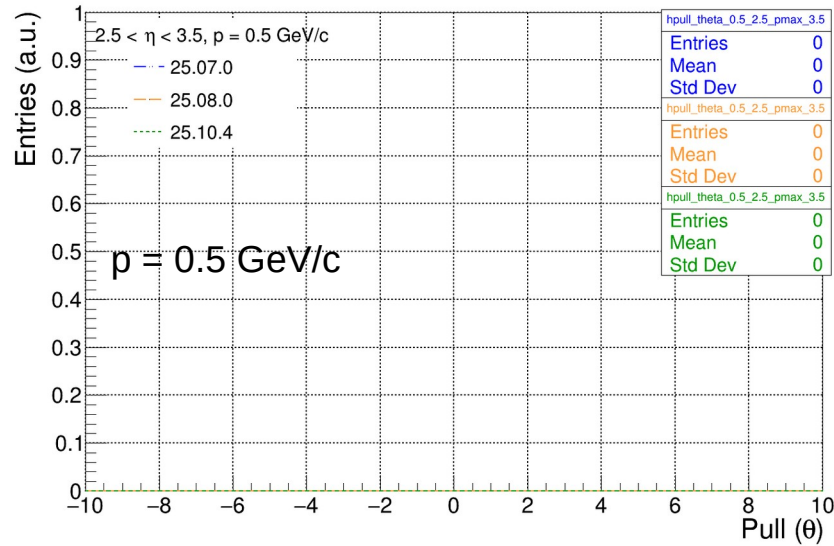


$-2.5 < \eta < -1.0$

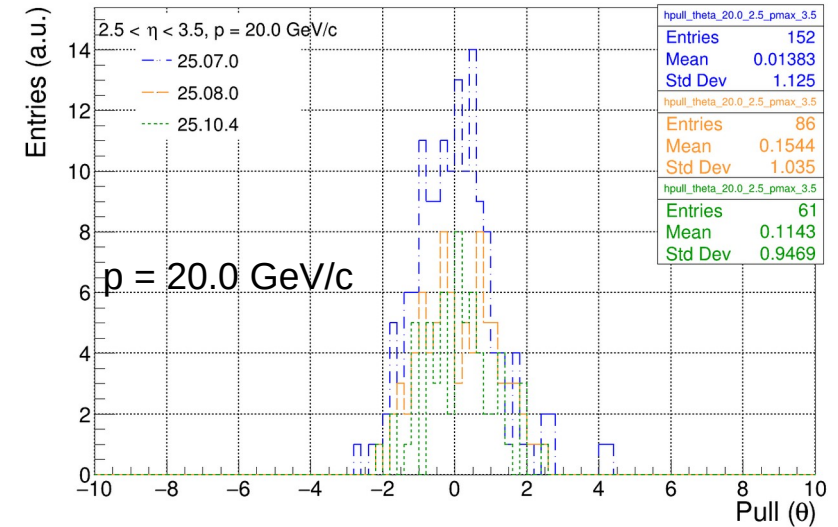
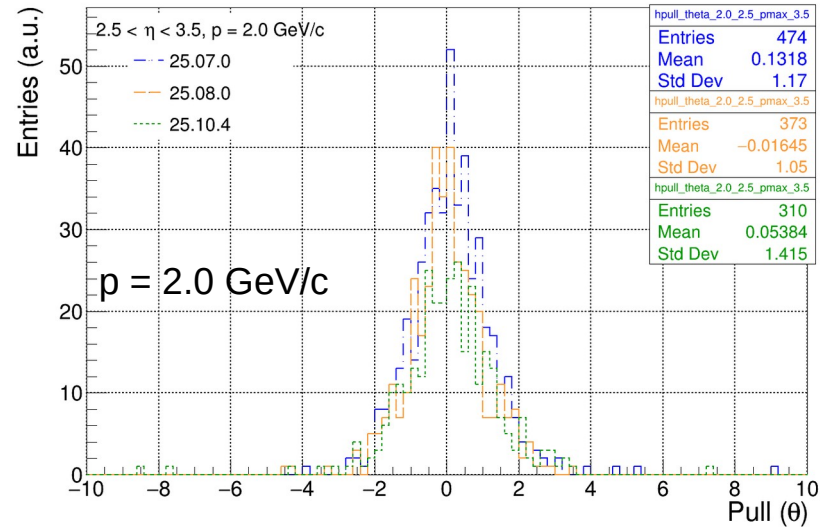
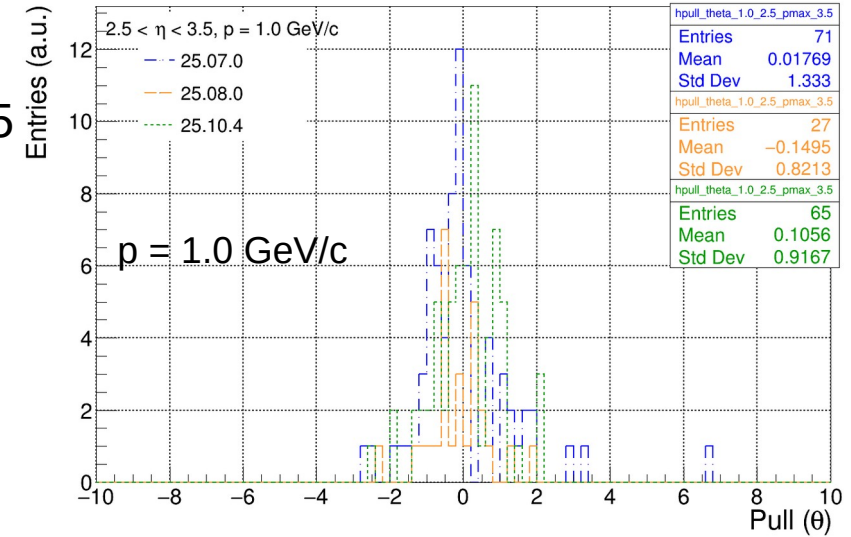




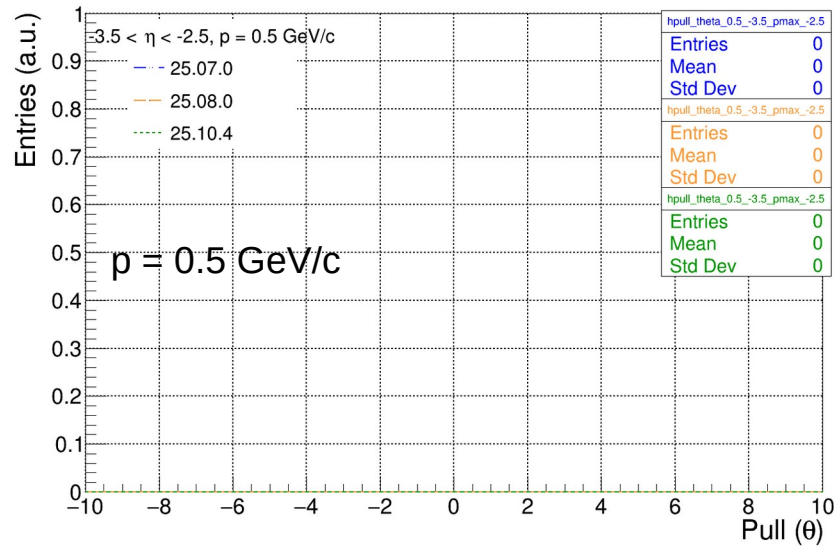
# Pull distributions ( $\theta$ )



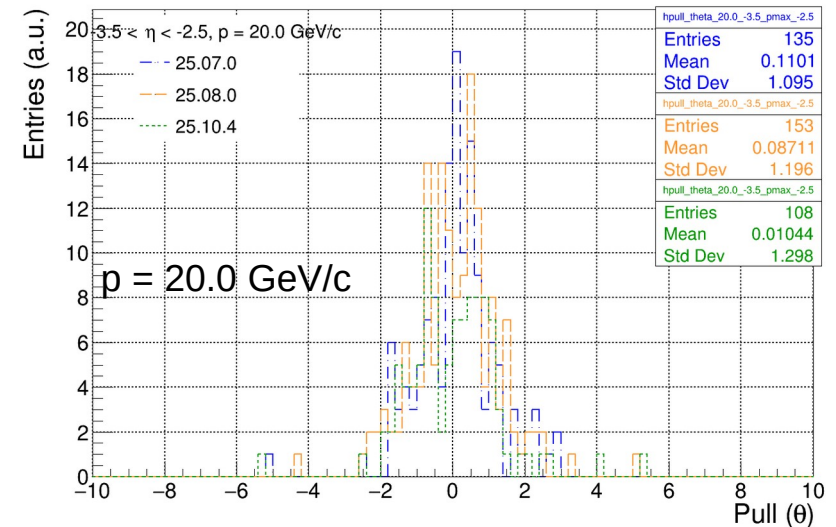
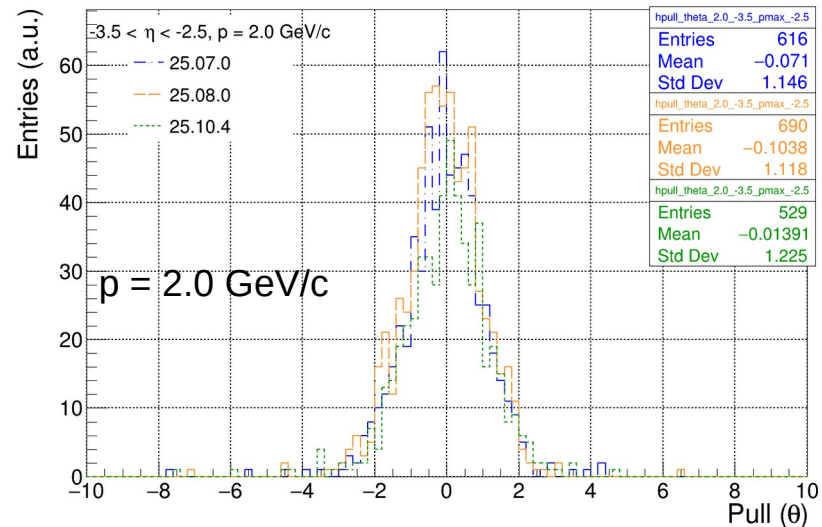
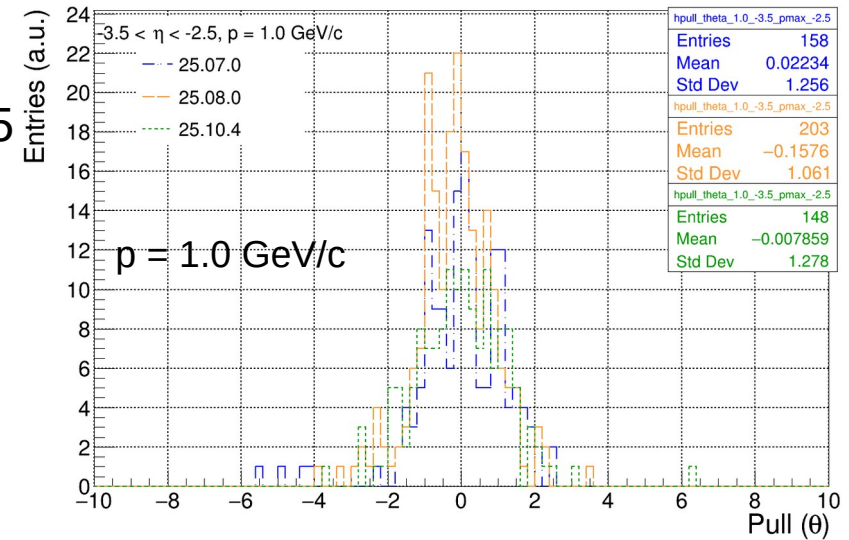
2.5 <  $\eta$  < 3.5



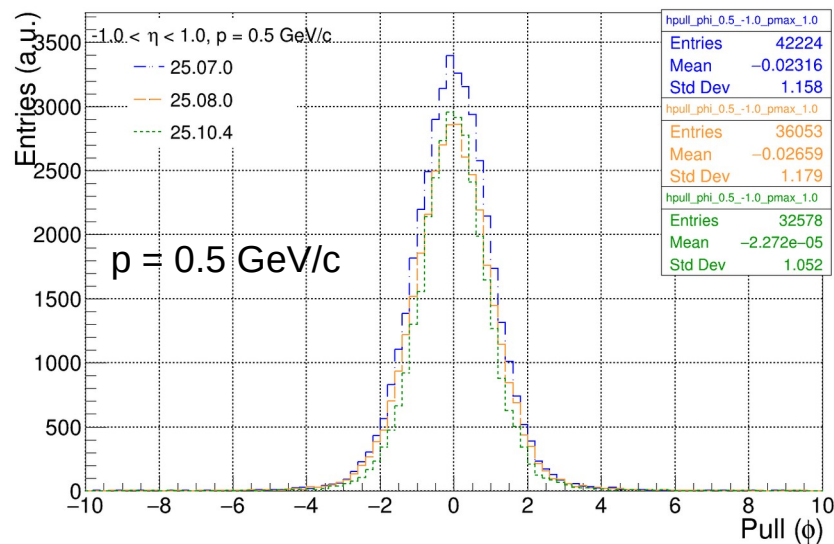
# Pull distributions ( $\theta$ )



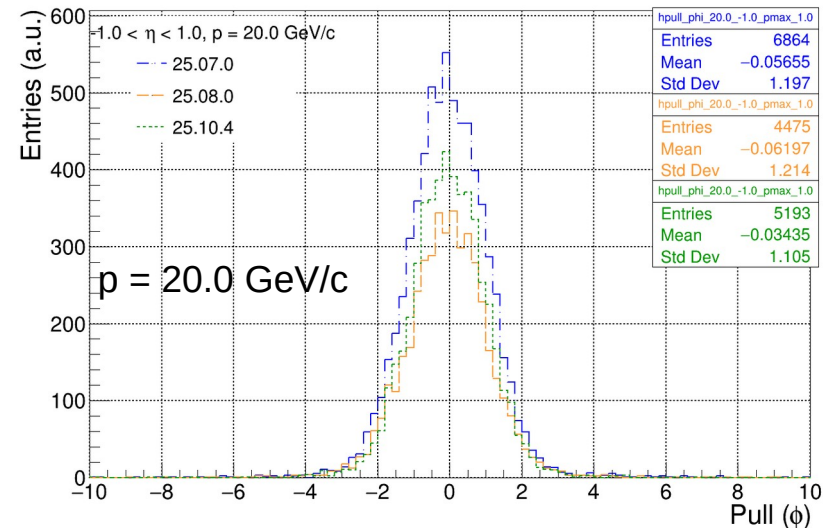
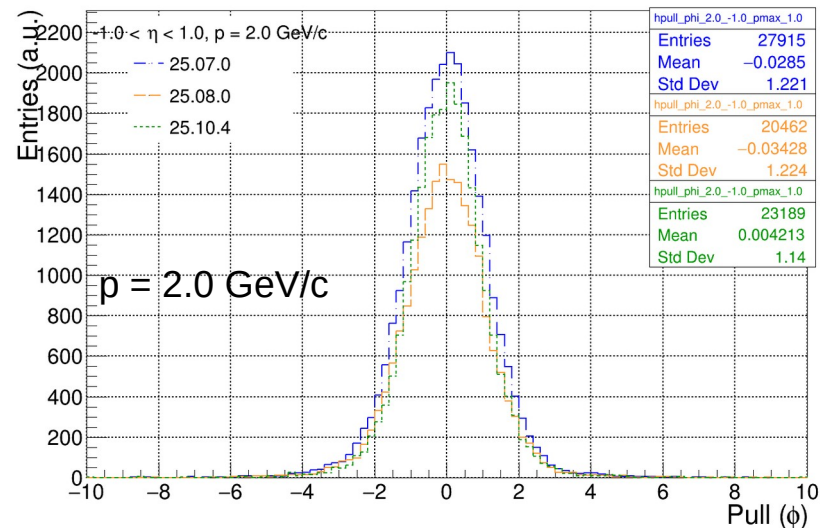
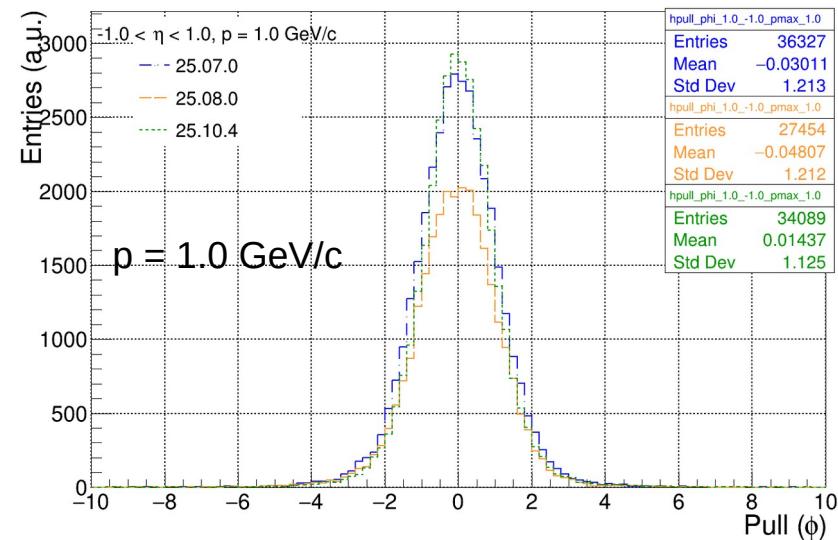
$-3.5 < \eta < -2.5$



# Pull distributions ( $\Phi$ )

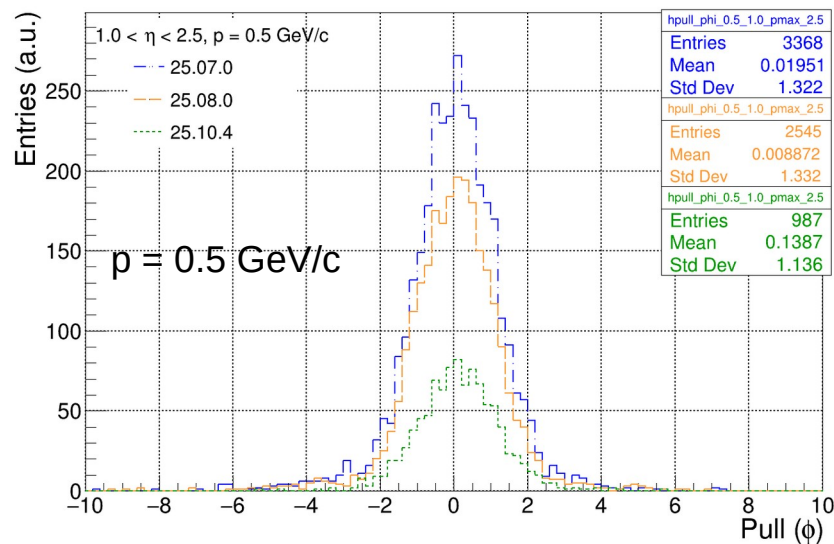


$-1 < \eta < 1$

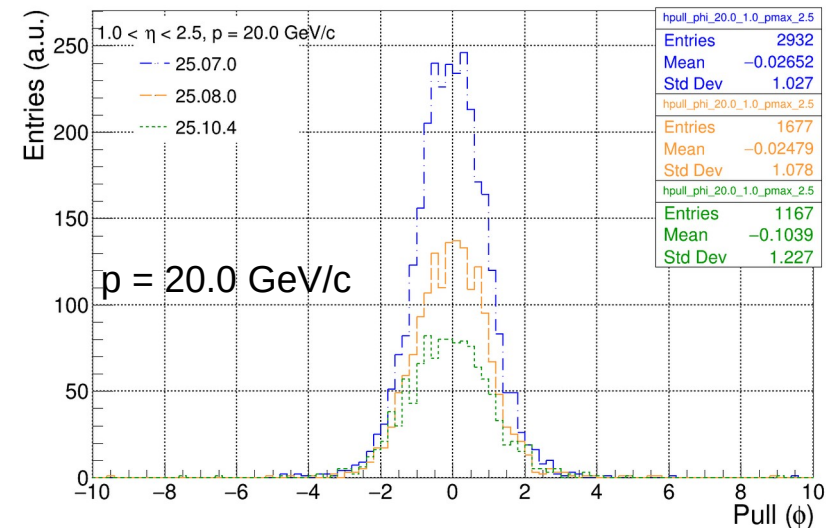
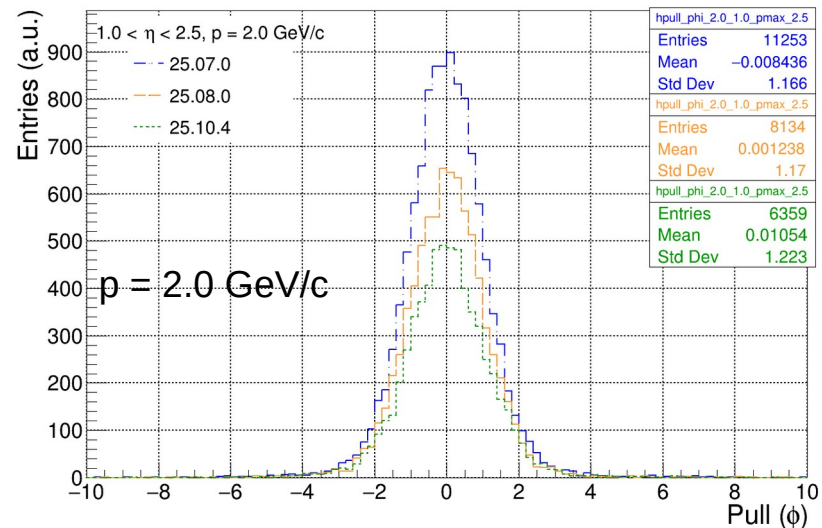
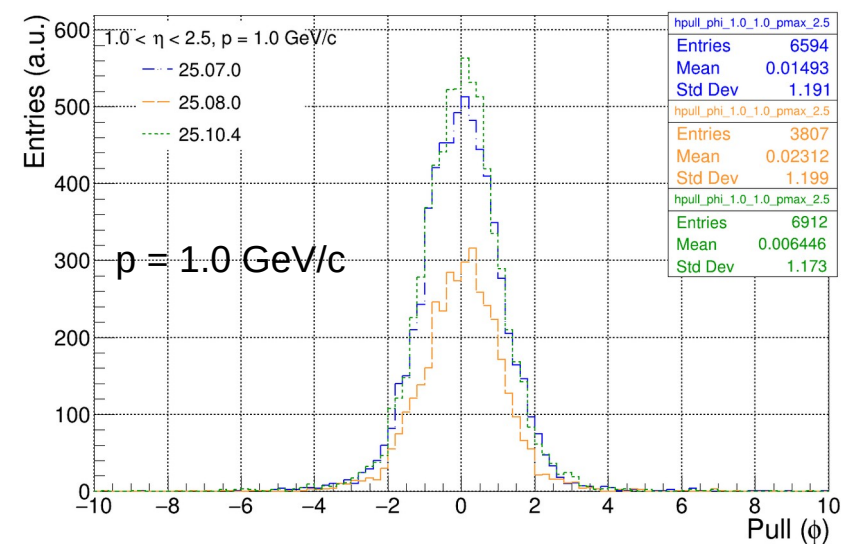




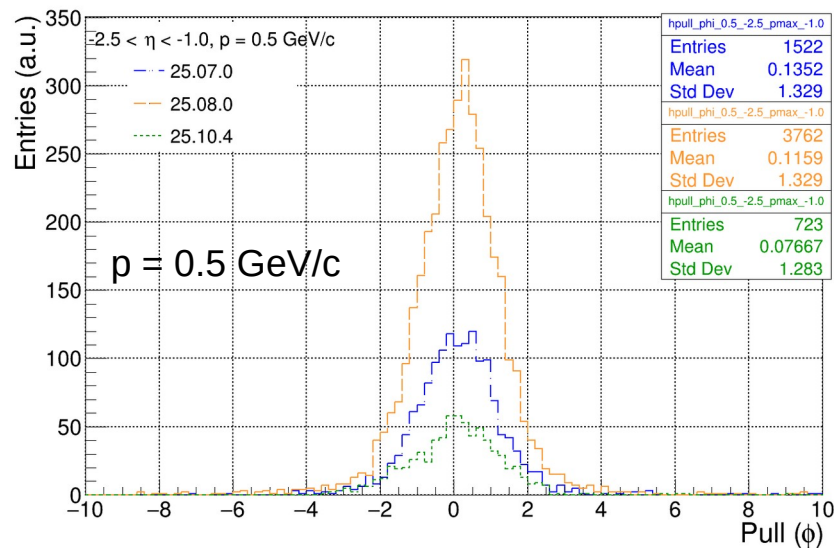
# Pull distributions ( $\Phi$ )



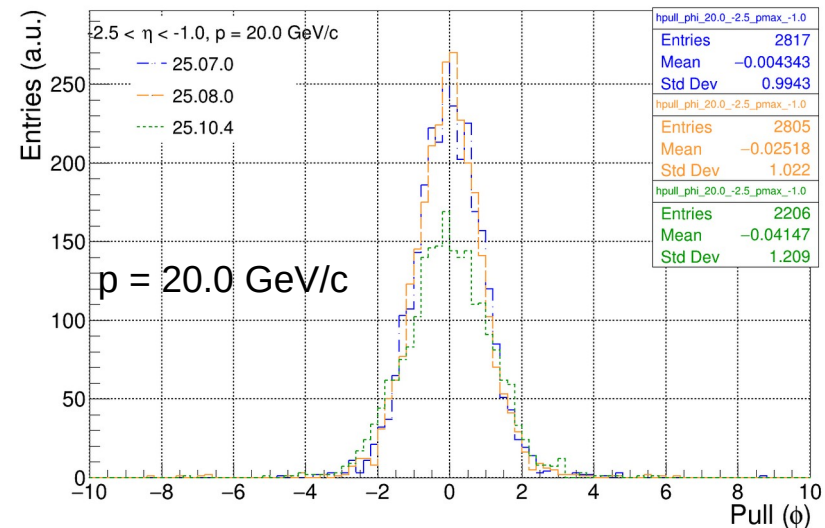
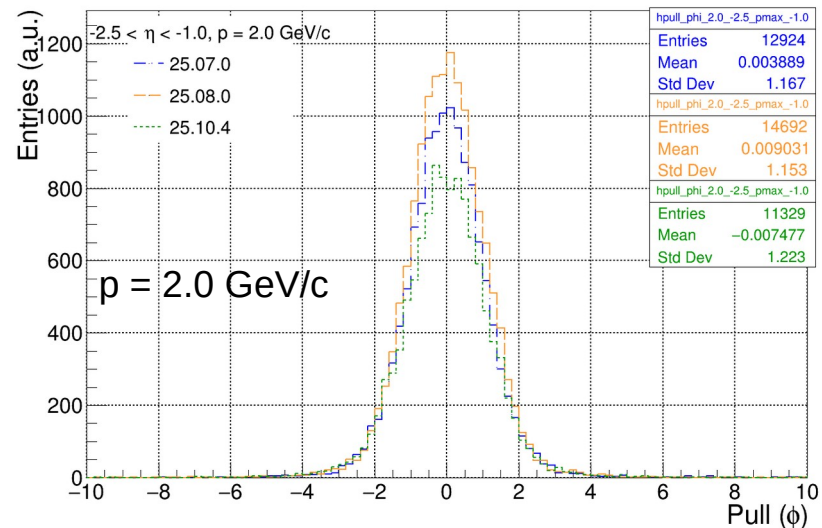
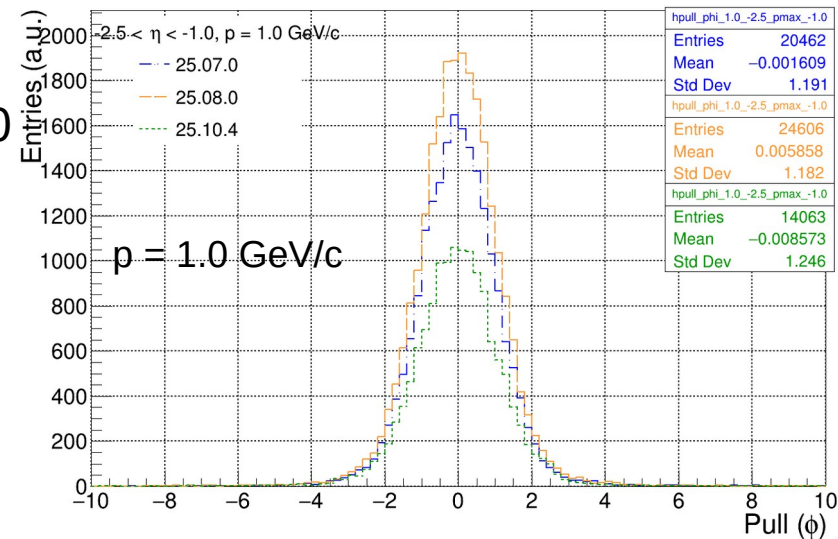
1.0 <  $\eta$  < 2.5



# Pull distributions ( $\Phi$ )

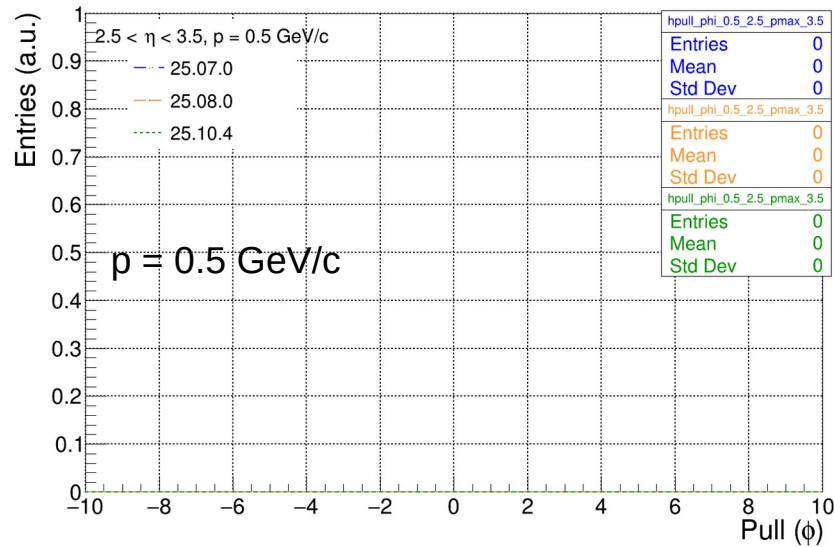


-2.5 <  $\eta$  < -1.0

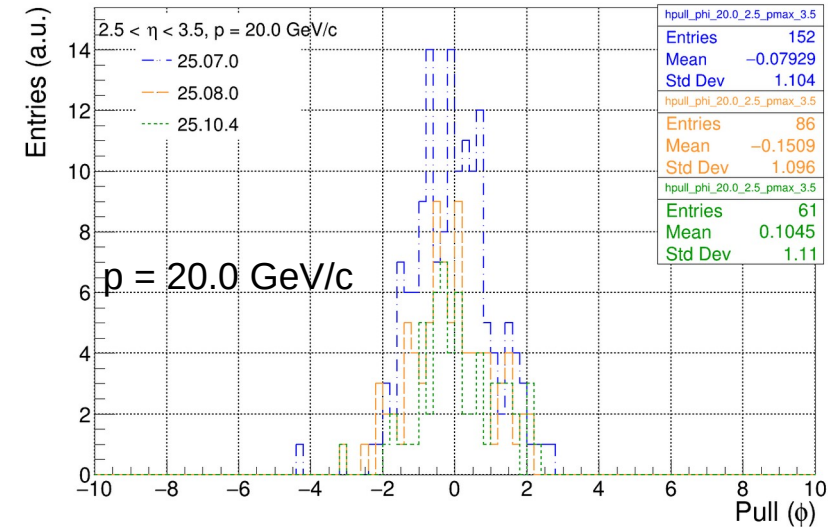
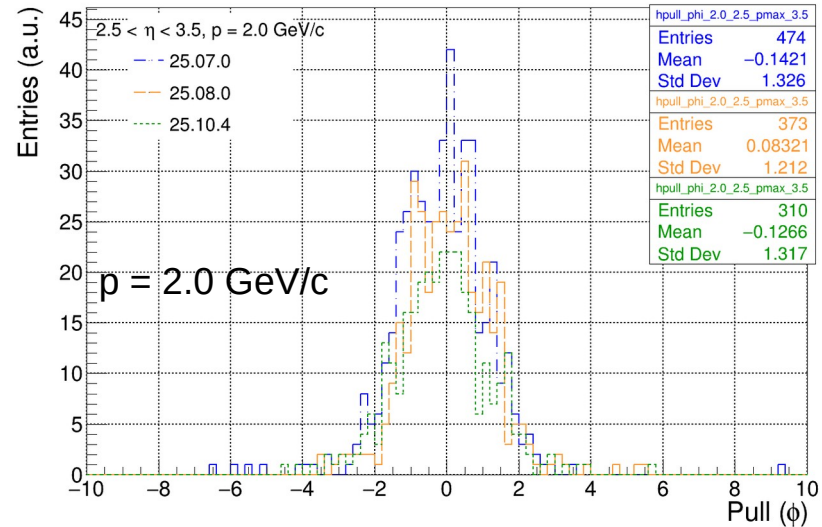
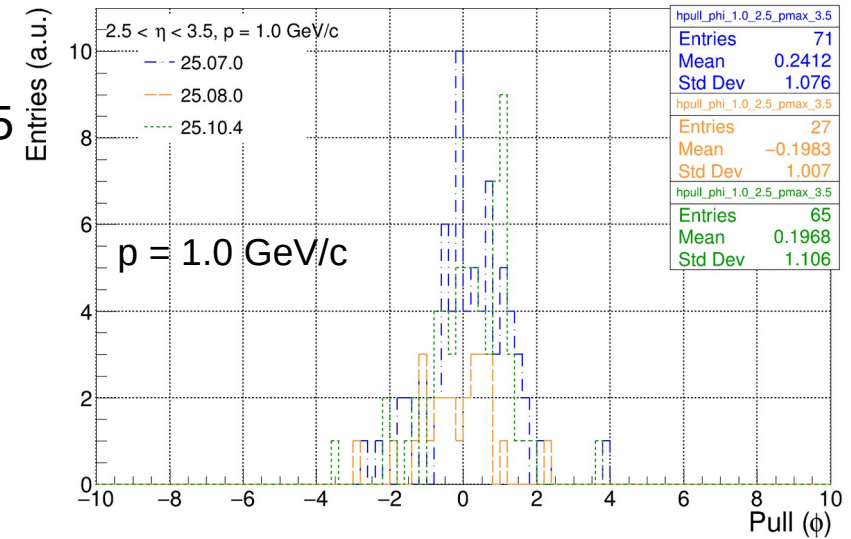




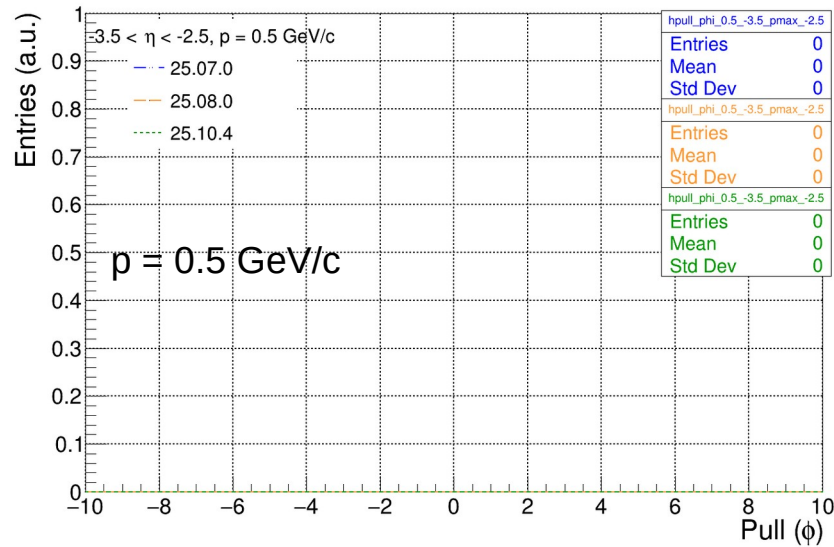
# Pull distributions ( $\Phi$ )



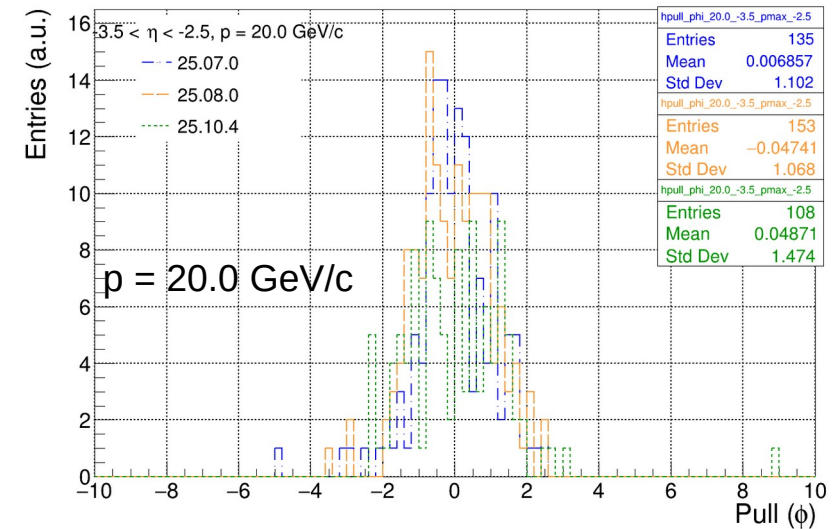
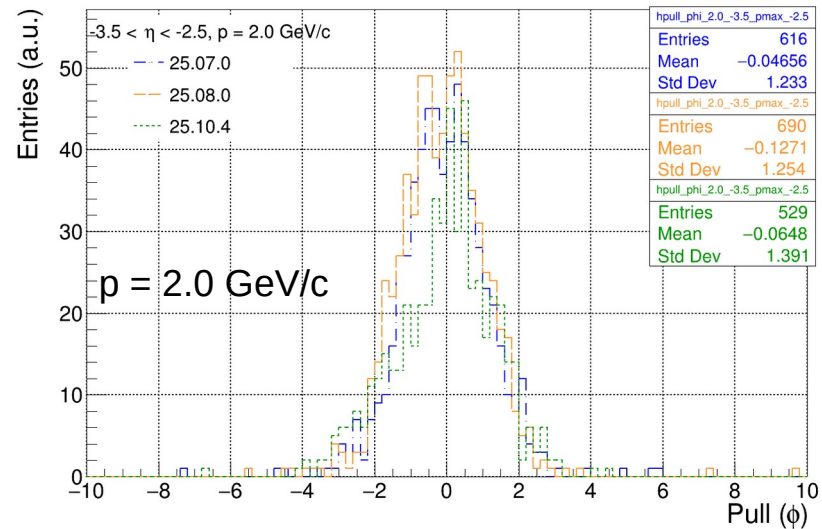
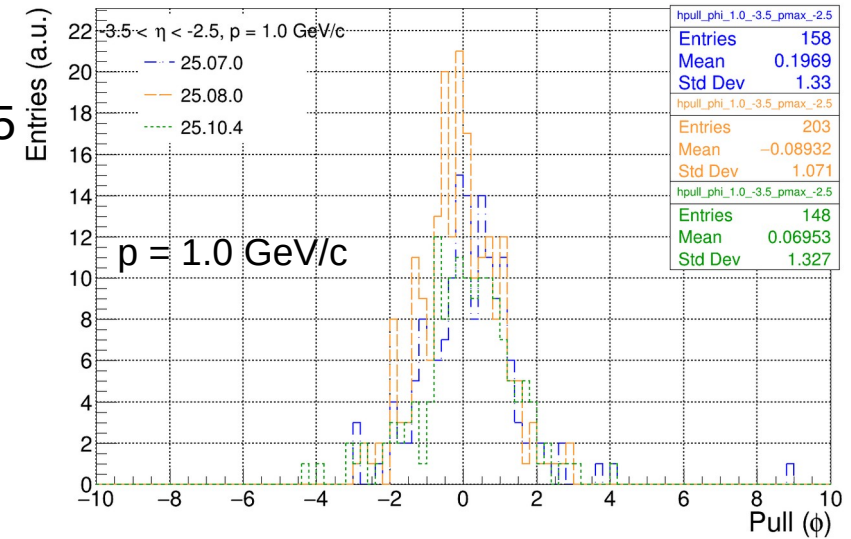
2.5 <  $\eta$  < 3.5



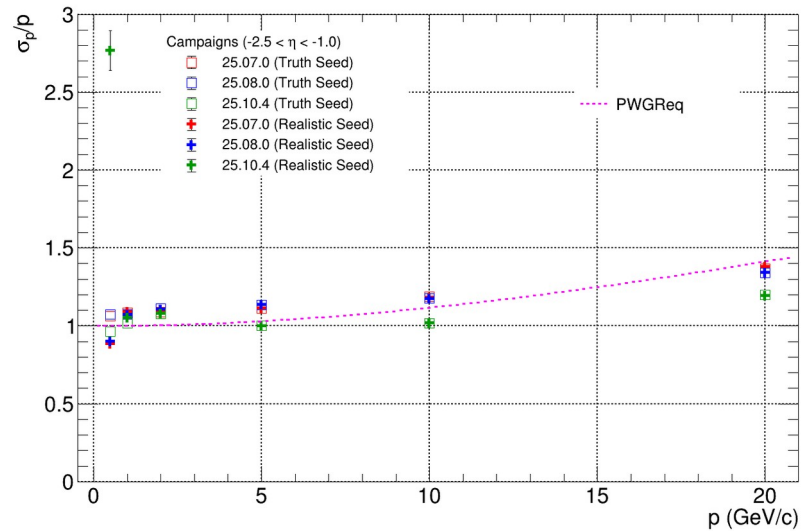
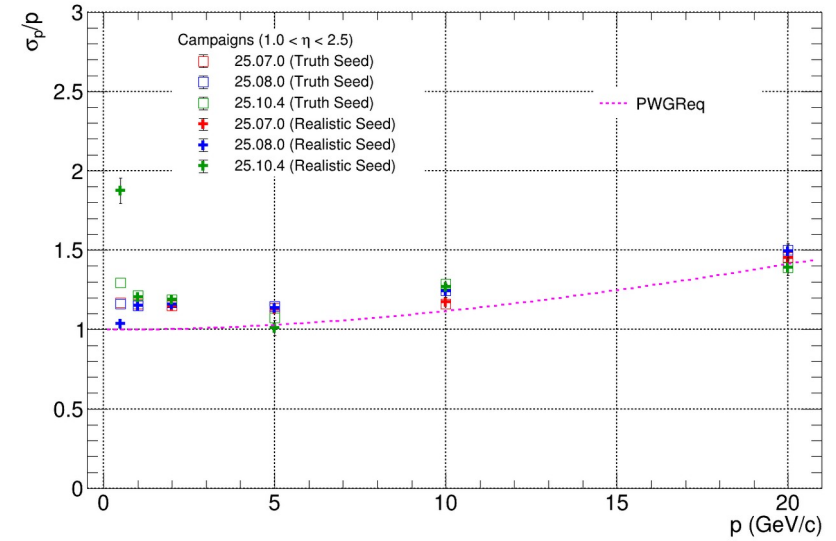
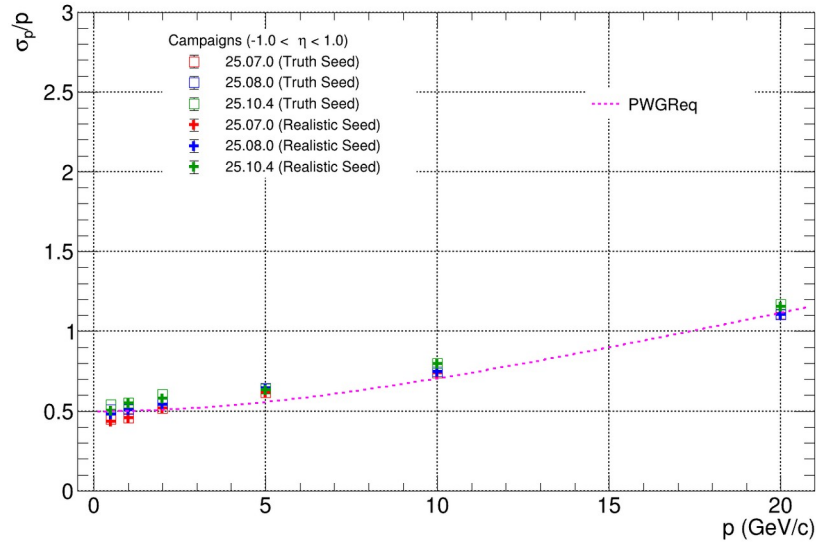
# Pull distributions ( $\Phi$ )



$-3.5 < \eta < -2.5$



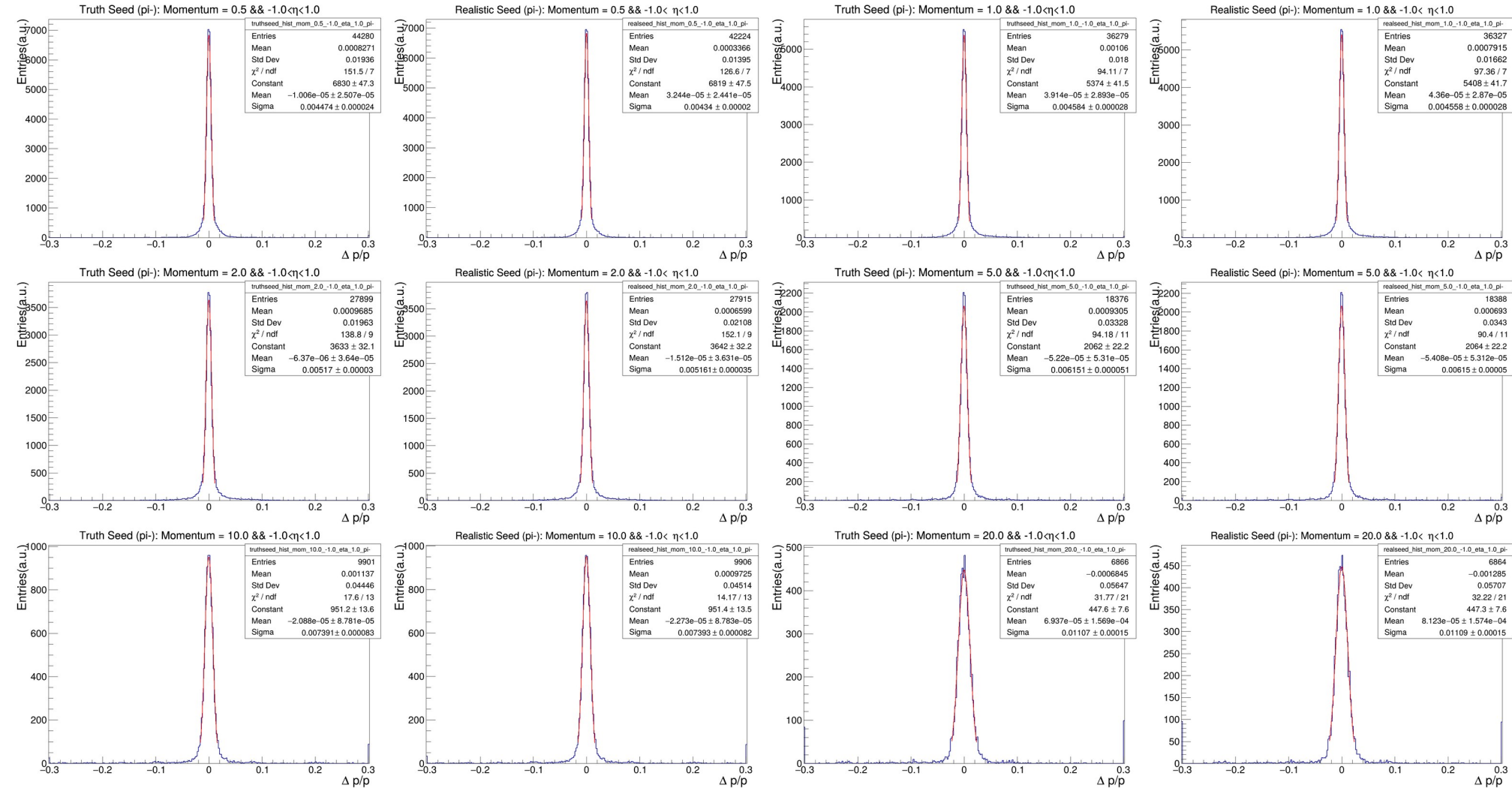
# Tracking Performances



Tracking performance based on Gaussian fit in (2 sigma) region

# Momentum debug plots (25.07.0)

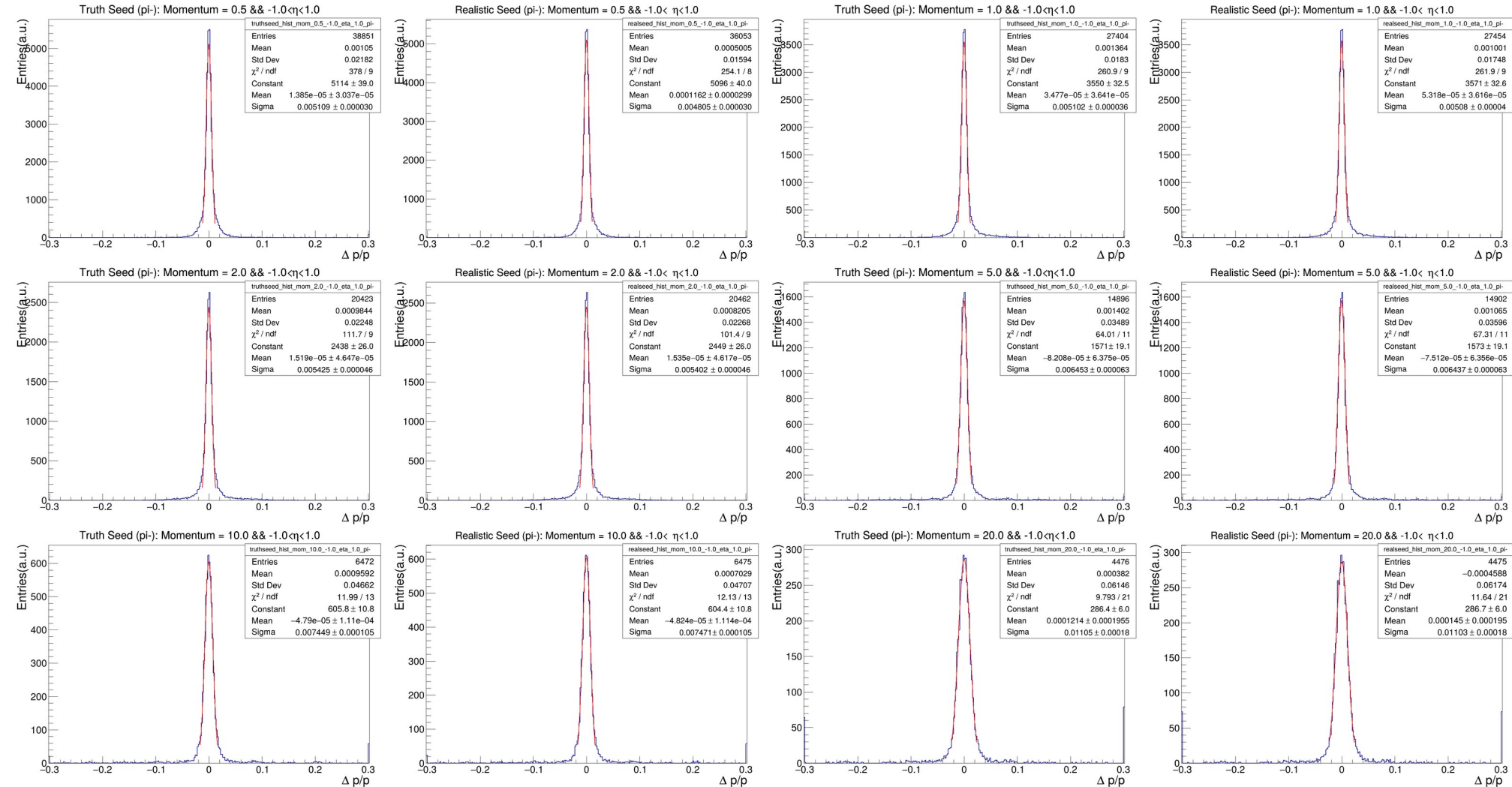
$$-1.0 < \eta < 1.0$$





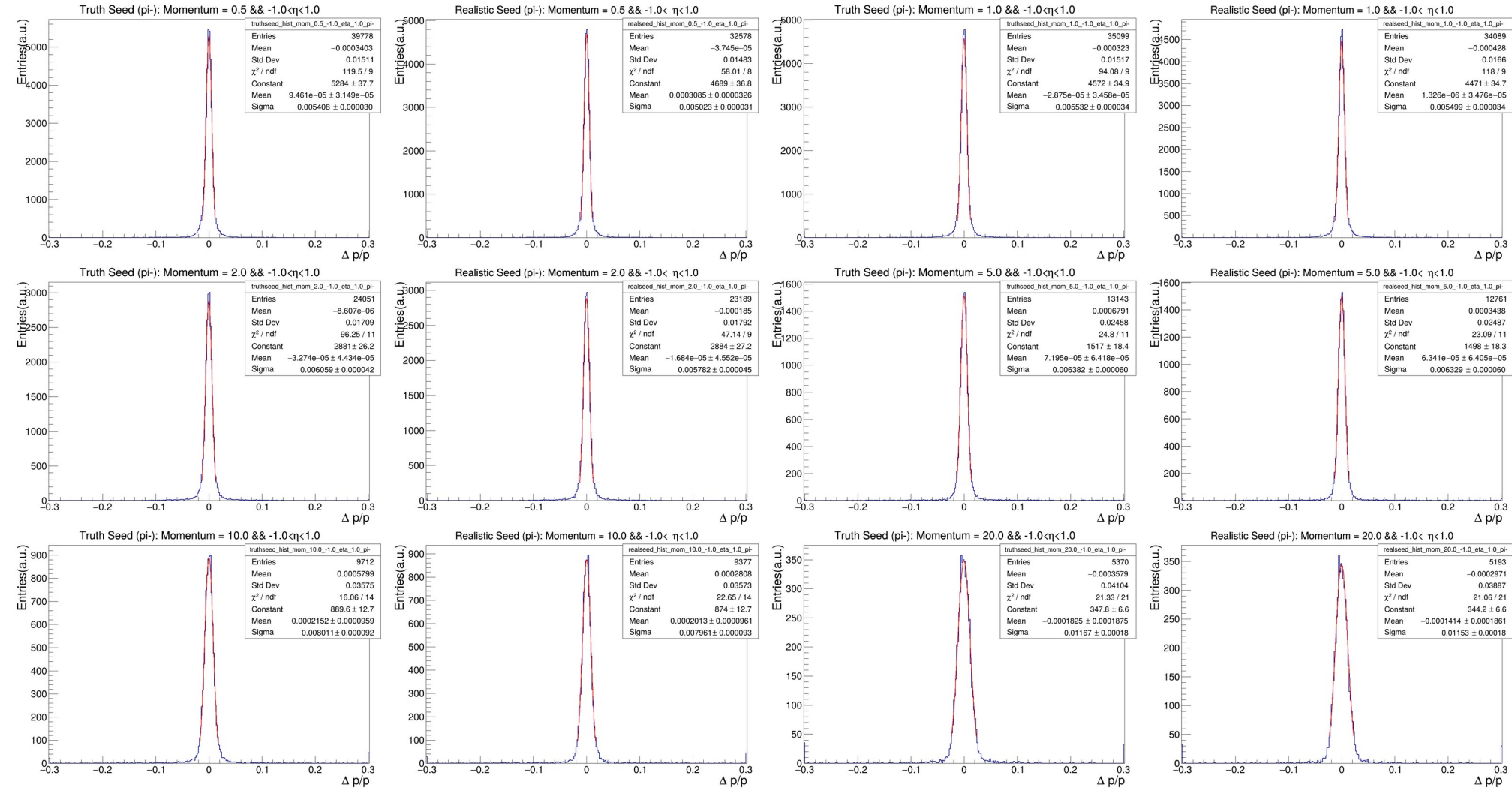
# Momentum debug plots (25.08.0)

$-1.0 < \eta < 1.0$



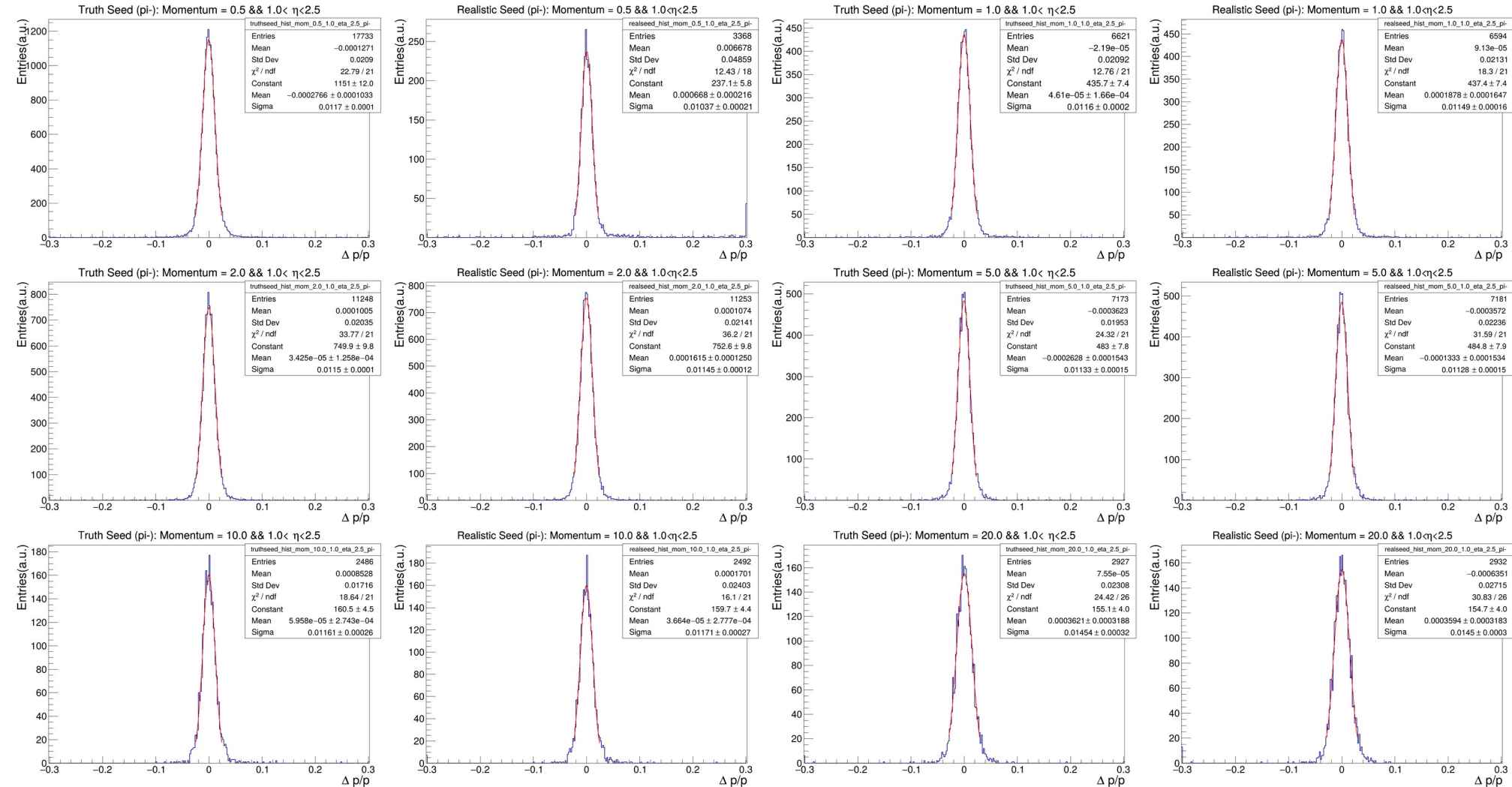
# Momentum debug plots (25.10.4)

$$-1.0 < \eta < 1.0$$



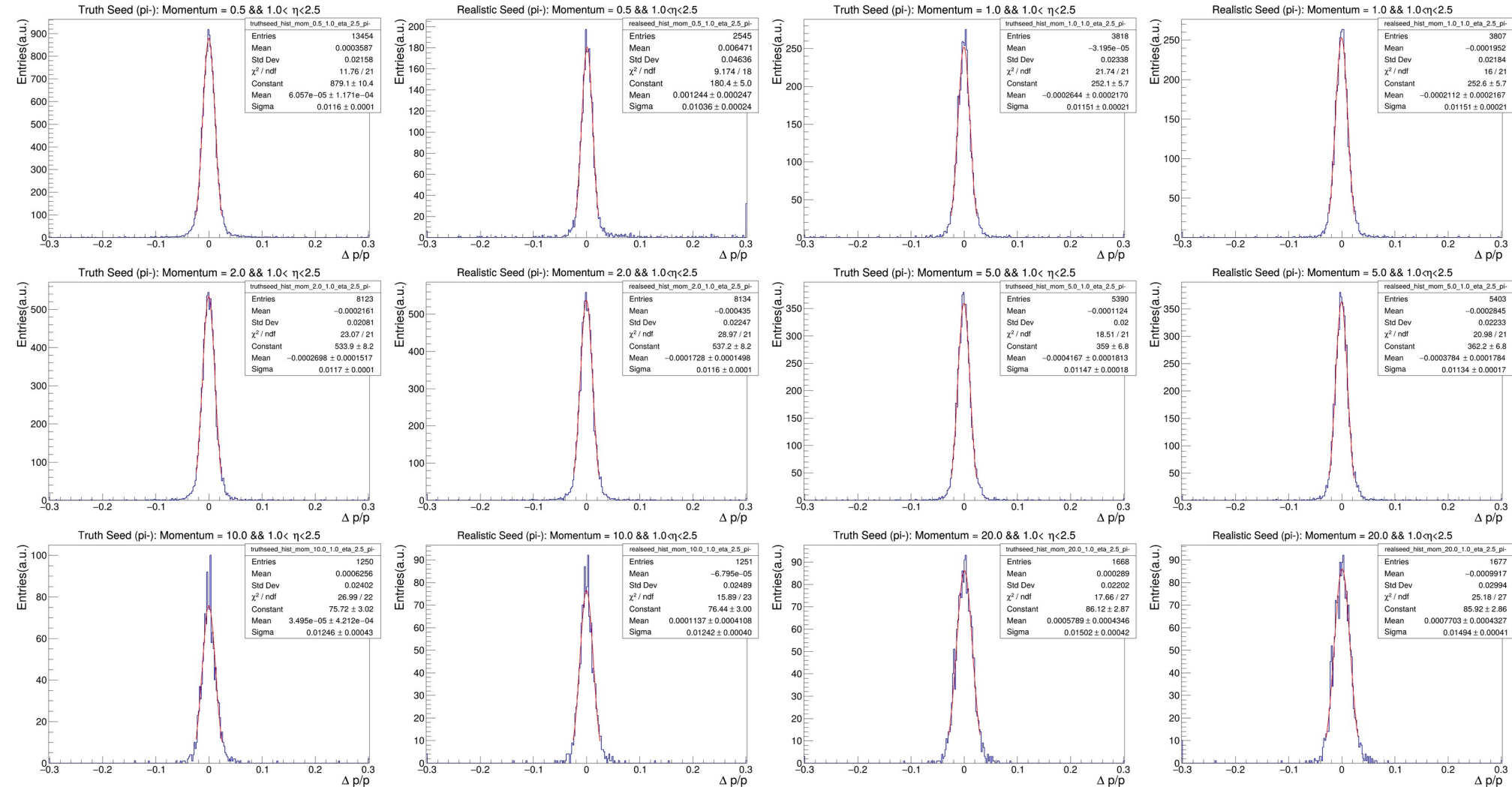
# Momentum debug plots (25.07.0)

$1.0 < \eta < 2.5$



# Momentum debug plots (25.08.0)

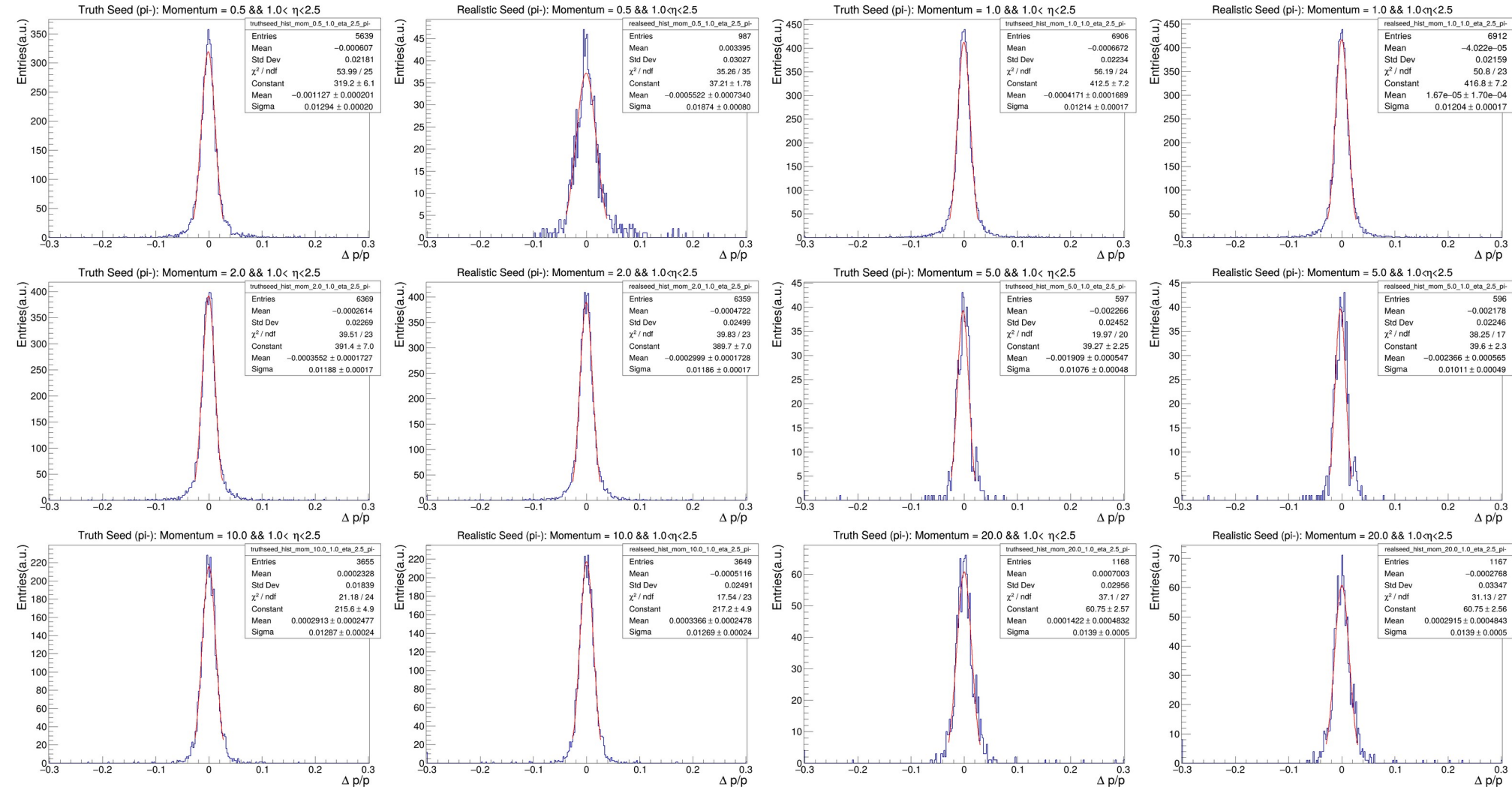
$$1.0 < \eta < 2.5$$





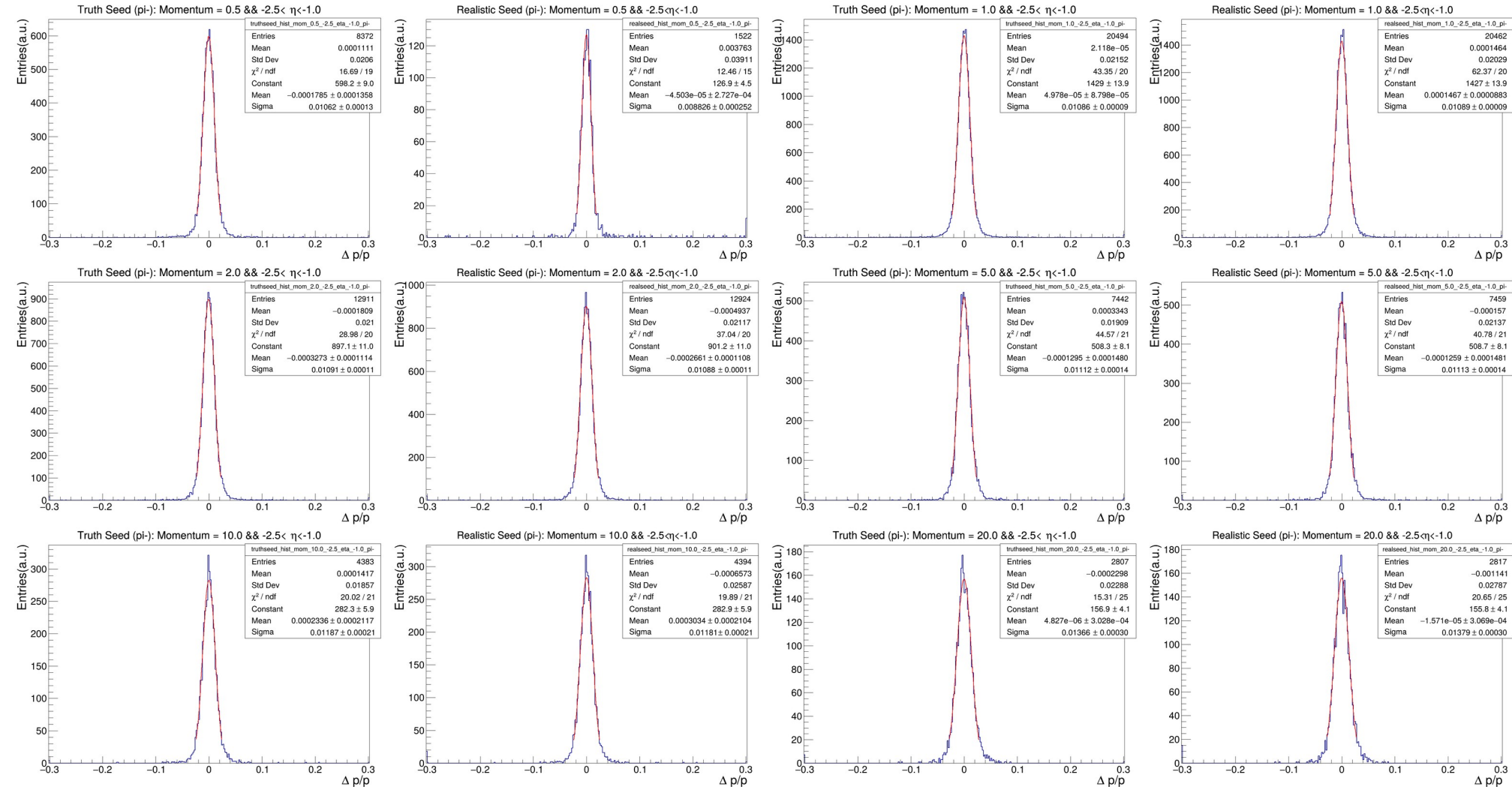
# Momentum debug plots (25.10.4)

$1.0 < \eta < 2.5$



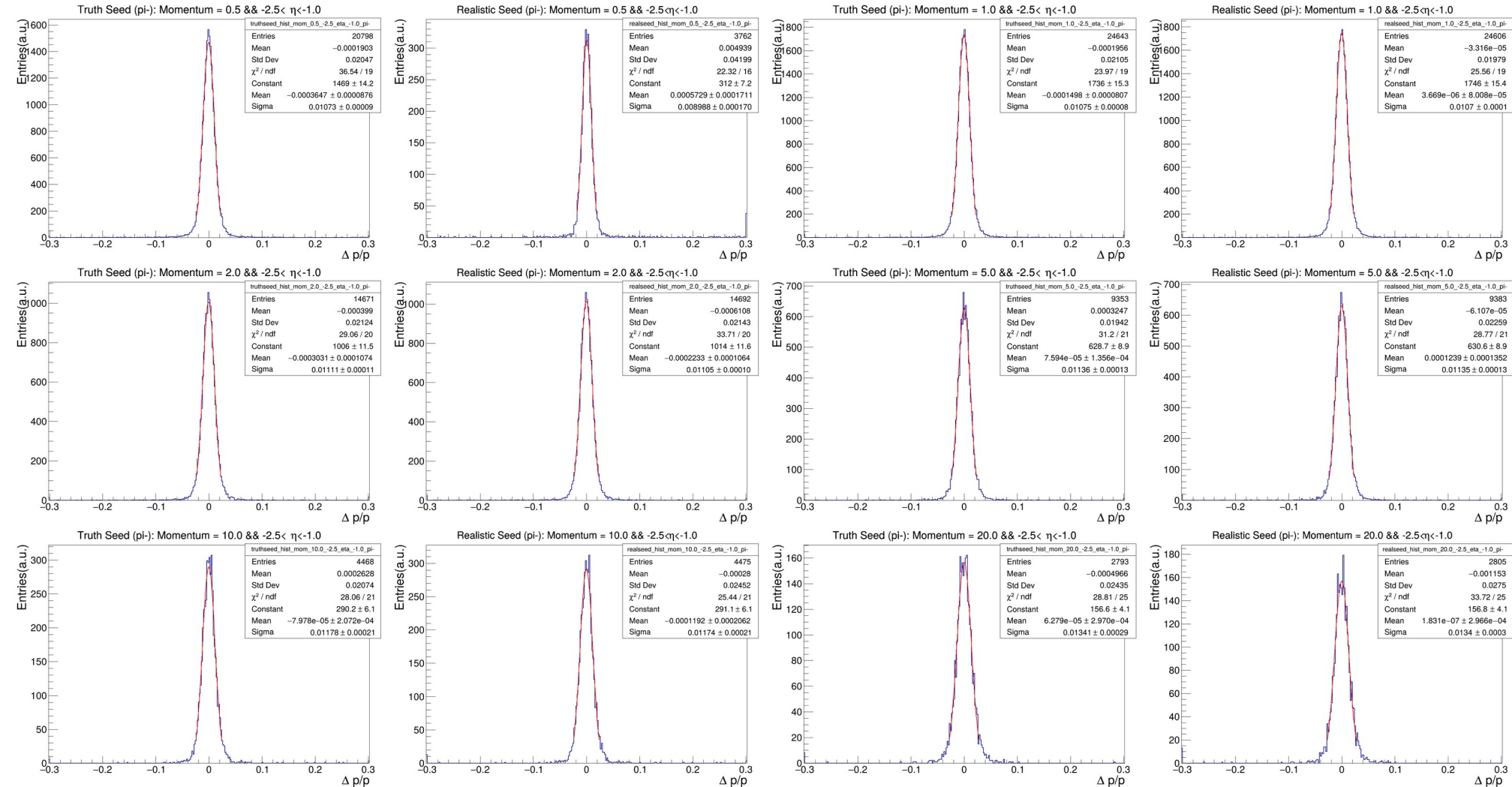
# Momentum debug plots (25.07.0)

$-2.5 < \eta < -1.0$



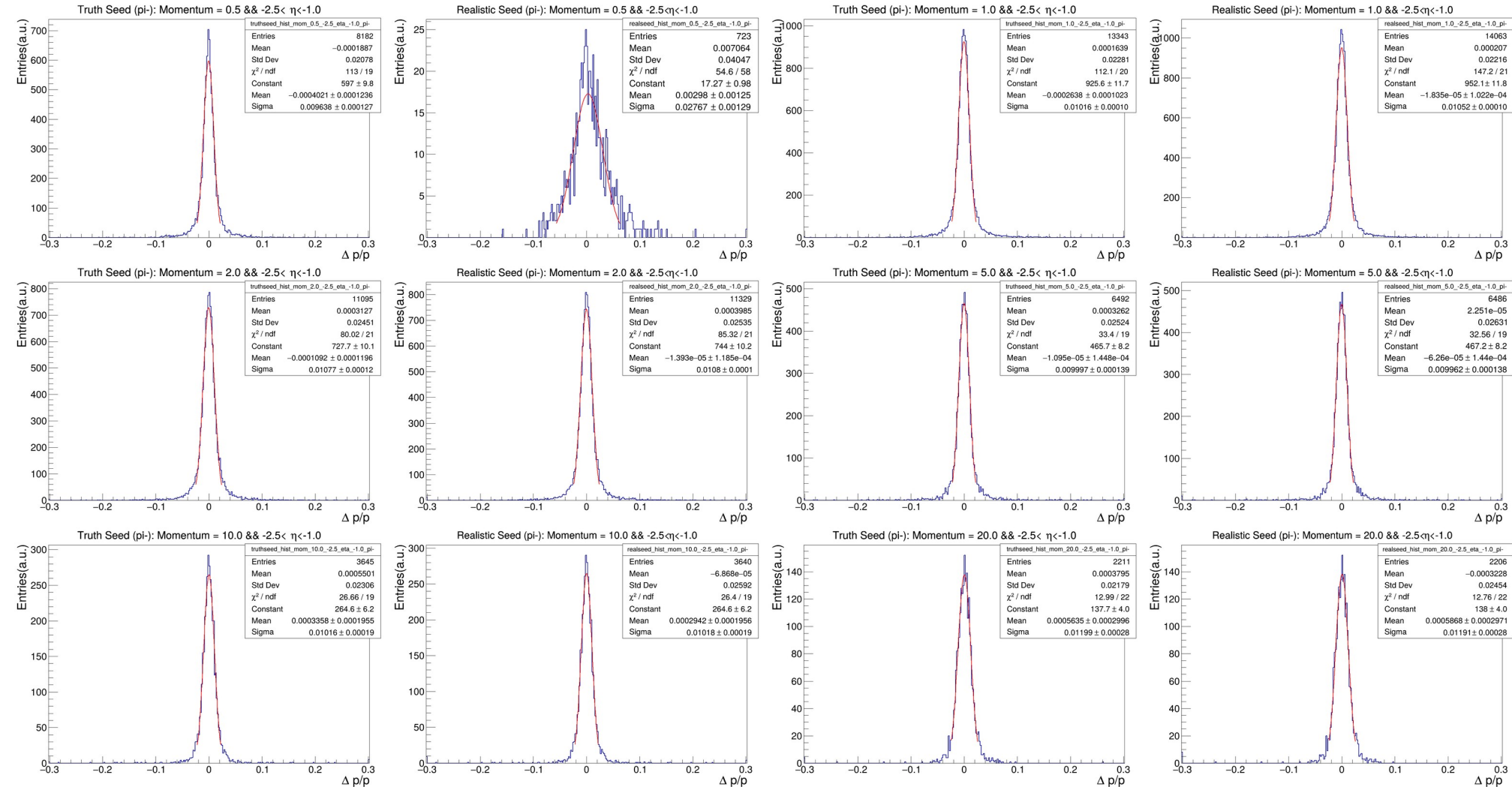
# Momentum debug plots (25.08.0)

$-2.5 < \eta < -1.0$



# Momentum debug plots (25.10.4)

$-2.5 < \eta < -1.0$





# Covariance Matrix in ACTS

For each fitted track we get track parameters and covariance matrix

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

	$l_0$	$l_1$	$\phi$	$\theta$	$q/p$	$\text{time}$
$l_0$	$\sigma^2(l_0)$	$\text{cov}(l_0, l_1)$	$\text{cov}(l_0, \phi)$	$\text{cov}(l_0, \theta)$	$\text{cov}(l_0, q/p)$	$\text{cov}(l_0, t)$
$l_1$	.	$\sigma^2(l_1)$	$\text{cov}(l_1, \phi)$	$\text{cov}(l_1, \theta)$	$\text{cov}(l_1, q/p)$	$\text{cov}(l_1, t)$
$\phi$	.	.	$\sigma^2(\phi)$	$\text{cov}(\phi, \theta)$	$\text{cov}(\phi, q/p)$	$\text{cov}(\phi, t)$
$\theta$	.	.	.	$\sigma^2(\theta)$	$\text{cov}(\theta, q/p)$	$\text{cov}(\theta, t)$
$q/p$	.	.	.	.	$\sigma^2(q/p)$	$\text{cov}(q/p, t)$
$\text{time}$	.	.	.	.	.	$\sigma^2(t)$

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[20] = 0.000333566
```

```
Cov[0] = cov(l0, l0)
Cov[1] = cov(l0, l1)
Cov[2] = cov(l1, l1)
Cov[3] = cov(l0, phi)
cov[4] = cov(l1, phi)
cov[5] = cov(phi, phi)
cov[6] = cov(l0, theta)
cov[7] = cov(l1, theta)
cov[8] = cov(phi, theta)
cov[9] = cov(theta, theta)
cov[10] = cov(l0, q/p)
```

```
Cov[11] = cov(l1, q/p)
Cov[12] = cov(phi, q/p)
Cov[13] = cov(theta, q/p)
Cov[14] = cov(q/p, q/p)
Cov[15] = cov(l0, time)
Cov[16] = cov(l1, time)
Cov[17] = cov(phi, time)
Cov[18] = cov(theta, time)
Cov[19] = cov(q/p, time)
Cov[20] = cov(time, time)
```

$$\sigma_{l_0} = \sqrt{\text{Cov}[0]} \quad \sigma_{l_1} = \sqrt{\text{Cov}[2]}$$

$$\sigma_{\phi} = \sqrt{\text{Cov}[5]} \quad \sigma_{\theta} = \sqrt{\text{Cov}[9]}$$

$$\sigma_{q/p} = \sqrt{\text{Cov}[14]}$$