



# KALMAN FILTER PERFORMANCE STUDY

## ND-GAR MEETING 19<sup>TH</sup> SEPTEMBER 2023



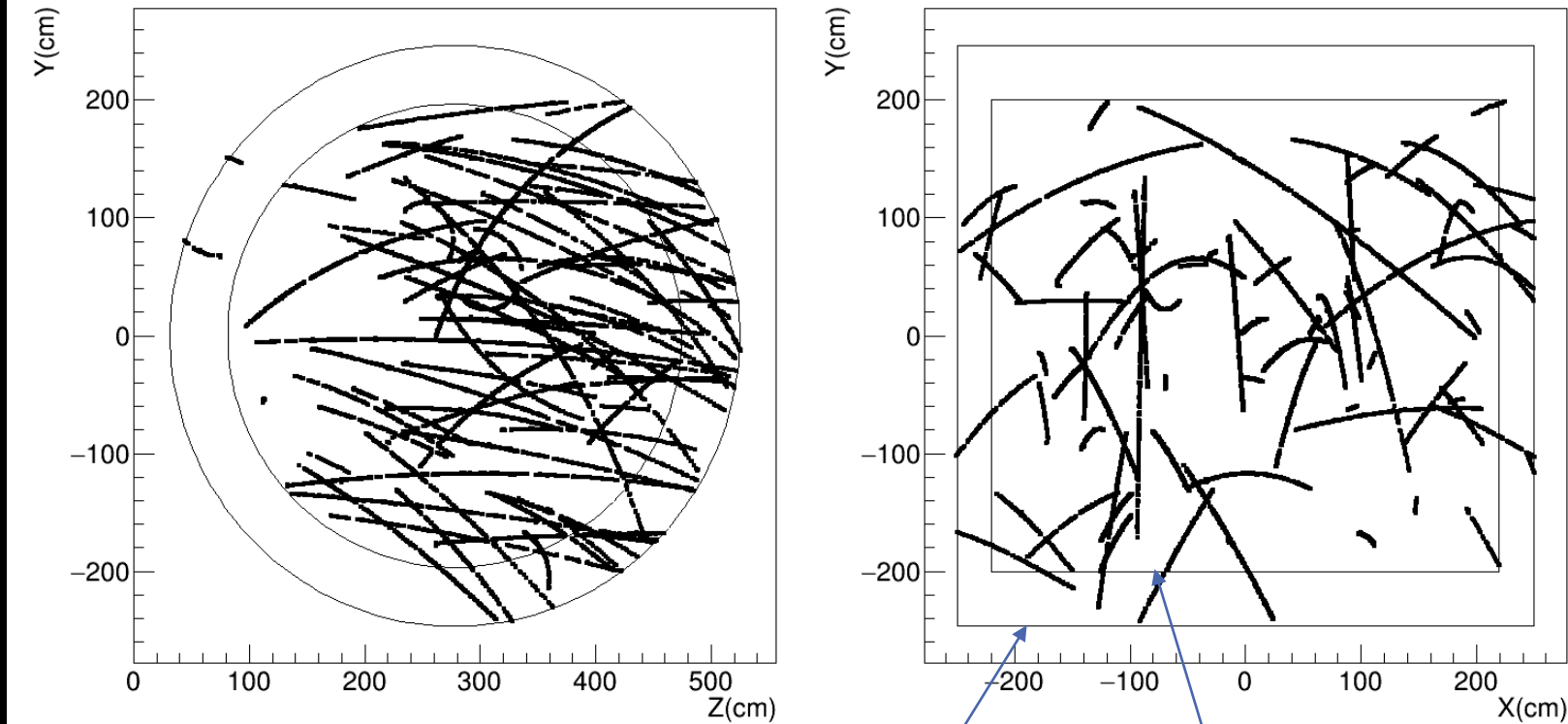
*Author:  
Federico Battisti*



# ALICE BASED KALMAN FILTER FOR ND-GAR: PERFORMANCE STUDY

- In today's presentation:
  - Study on sample of primary particles ( $\mu^-$ ,  $p$ ,  $\pi^+$ ) produced in  $\nu_\mu CC$  interactions inside the TPC fiducial volume:
    - Discussion of bug fixes for new ALICE-based Kalman Filter
    - Comparison of momentum reconstruction performance of new ALICE-based Kalman Filter with current GArSoft reconstruction
- Previous presentations include:
  1. Dune Collaboration meeting 26<sup>th</sup> January 2022: <https://indico.fnal.gov/event/50215/contributions/232480/>
  2. ND-GAr weekly meeting 15<sup>th</sup> March 2022: <https://indico.fnal.gov/event/53600/contributions/236685/>
  3. DUNE Collaboration meeting 18<sup>th</sup> May 2022: <https://indico.fnal.gov/event/50217/contributions/241519/>
  4. ND-GAr weekly meeting 9<sup>th</sup> August 2022: <https://indico.fnal.gov/event/55842/>
  5. ND-GAr weekly meeting 25<sup>th</sup> October 2022: <https://indico.fnal.gov/event/56687/>
  6. ND-GAr weekly meeting 28<sup>th</sup> February 2023: <https://indico.fnal.gov/event/58350/>
  7. Dune Collaboration meeting 25<sup>th</sup> May 2023: <https://indico.fnal.gov/event/57487/contributions/267579/>

## SAMPLE 2: MUONS FROM $\nu_{\mu}CC$ IN TPC FIDUCIAL

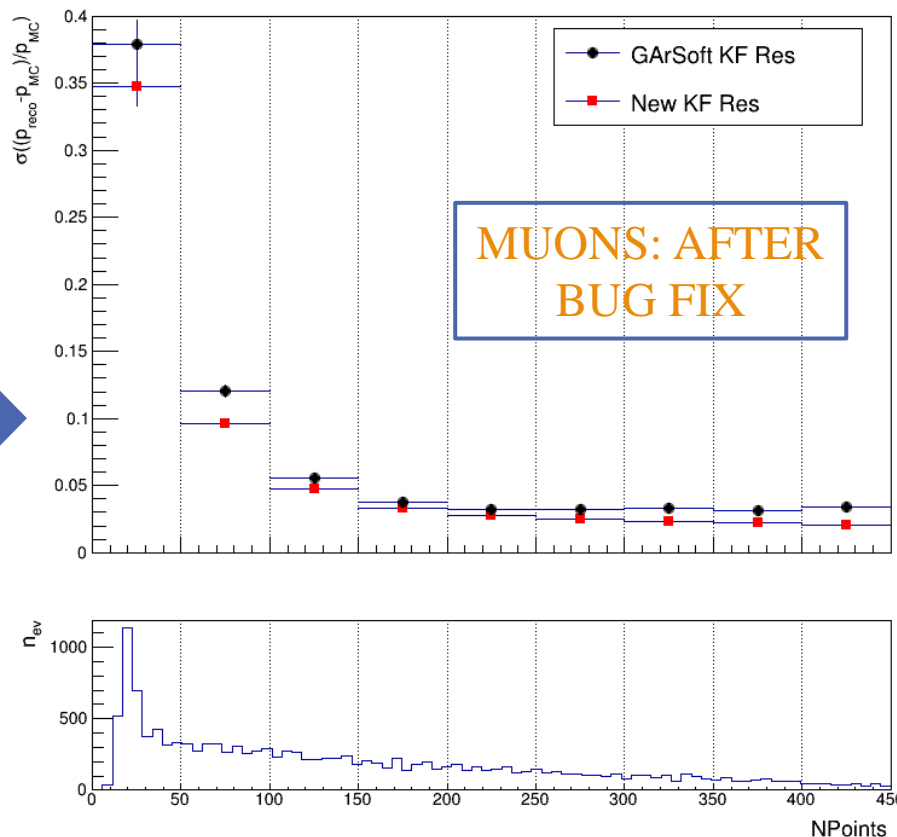
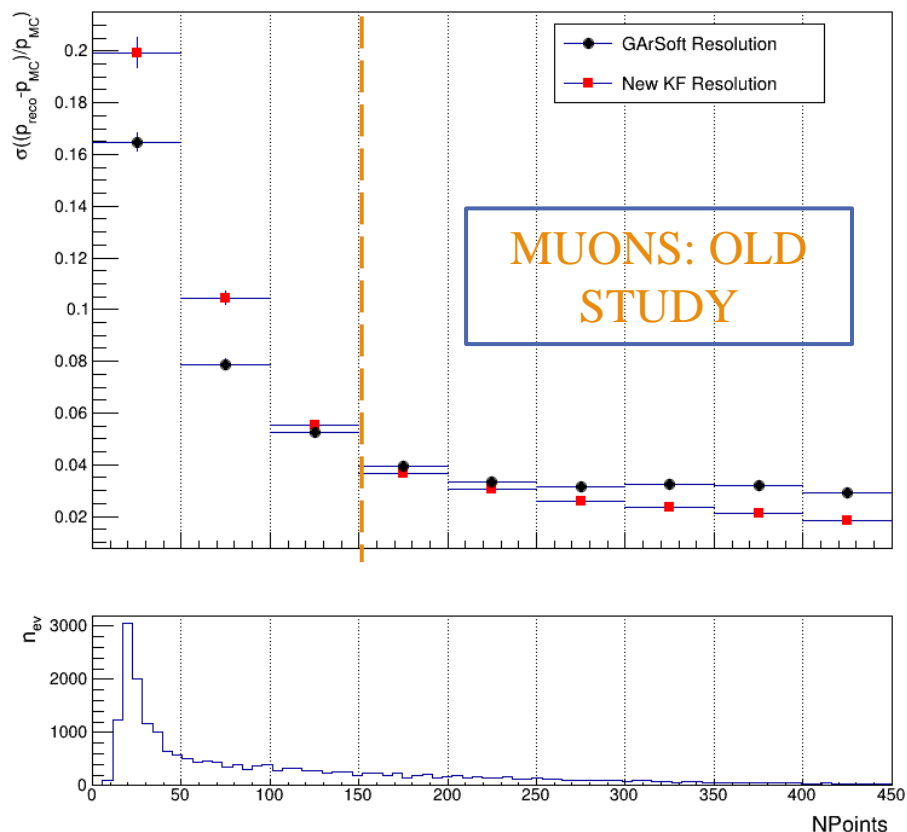


Total TPC Volume  
 $R_{TPC} \leq 246.6cm;$   
 $|z_{TPC}| \leq 249.6cm;$

Fiducial Volume  
 $R_{fid} \leq (R_{TPC} - 50cm);$   
 $|z_{fid}| \leq (|z_{TPC}| - 30cm);$

- **SAMPLE** :  $4.35 \times 10^4$  neutrino interactions in active TPC volume produced using GENIE module in GArSoft v2\_18\_00 with standard flux
- Selected only  $\nu_{\mu}CC$  interactions with reconstructed vertex in TPC fiducial volume as defined in ND-CDR :  
$$R_{fid} \leq (R_{TPC} - 50cm);$$
$$|z_{fid}| \leq (|z_{TPC}| - 30cm);$$
- Considered primary particles from interactions:  $\mu^{-}, p, \pi^{+}$  (previous study only included muons)

# NEW RESULTS AFTER BUG FIXES

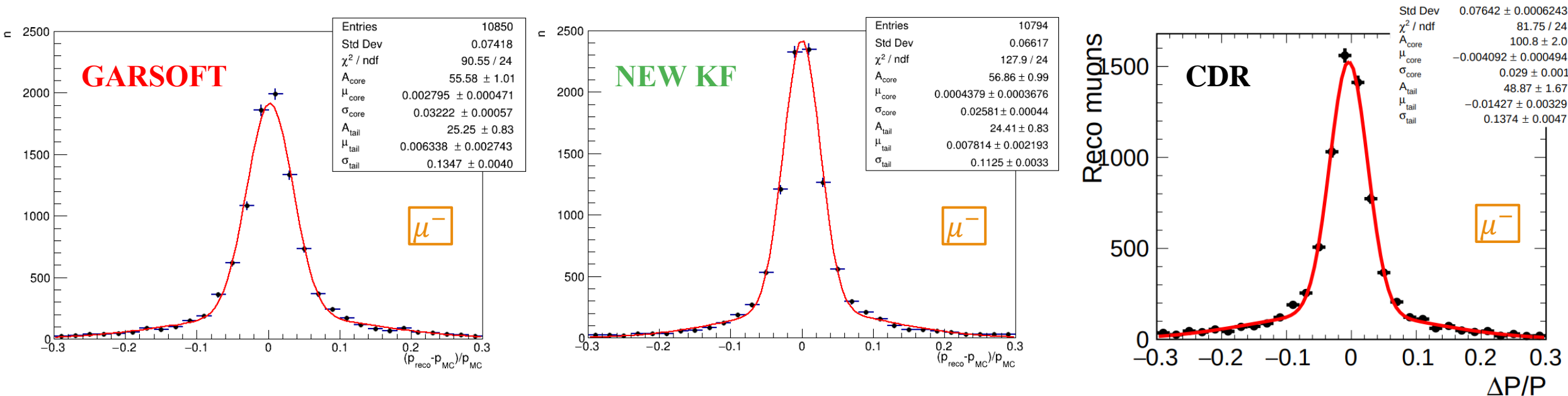


Profile plots for resolution : ( $\sigma$ ) from momentum residual Gauss fit in each NPoints slice

NPoints 1D distribution (NB: in old study garsoft tracks associated with wrong number of points)

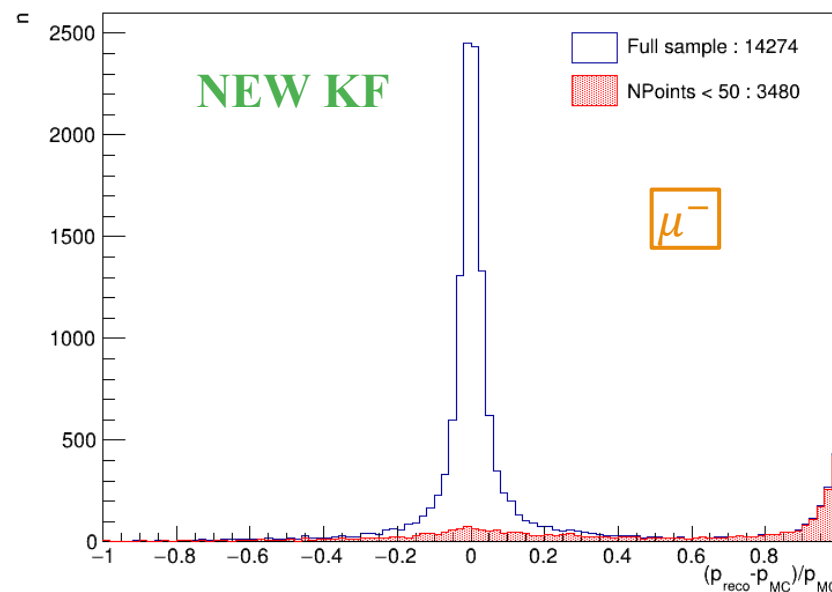
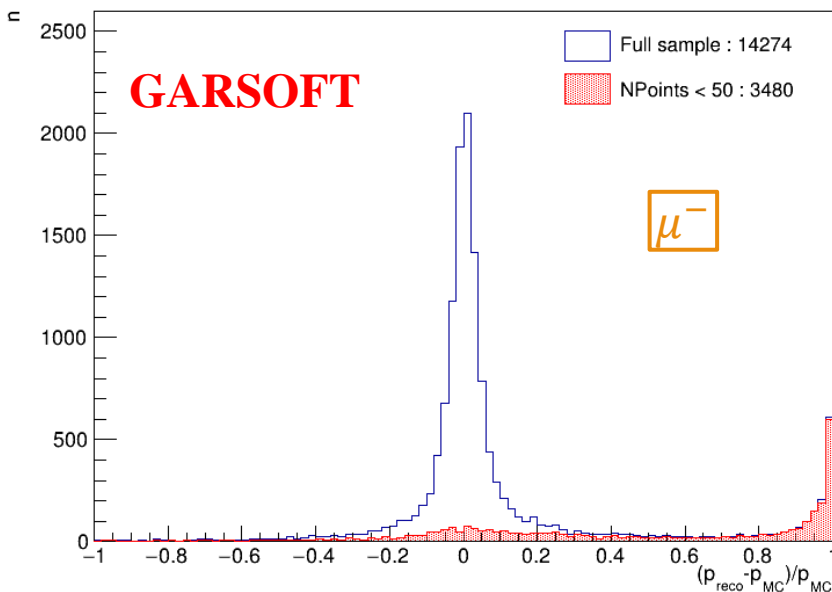
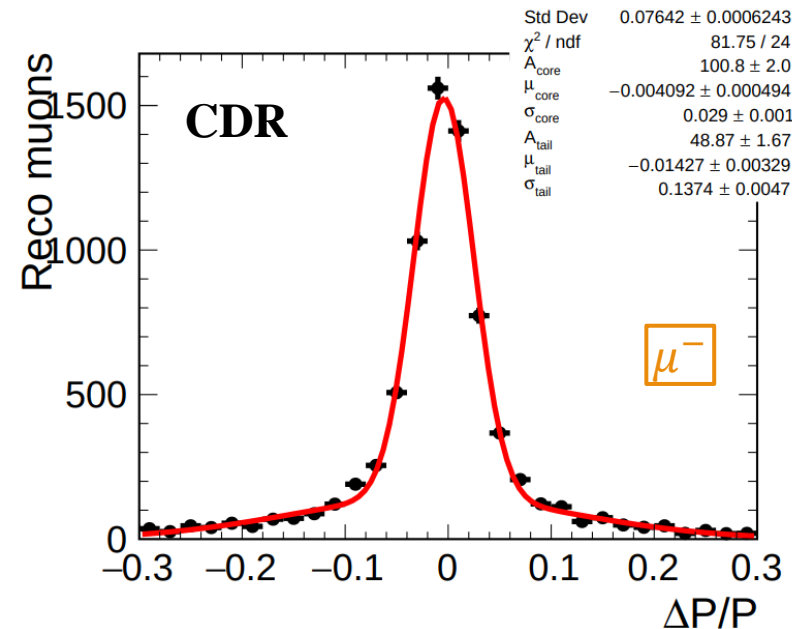
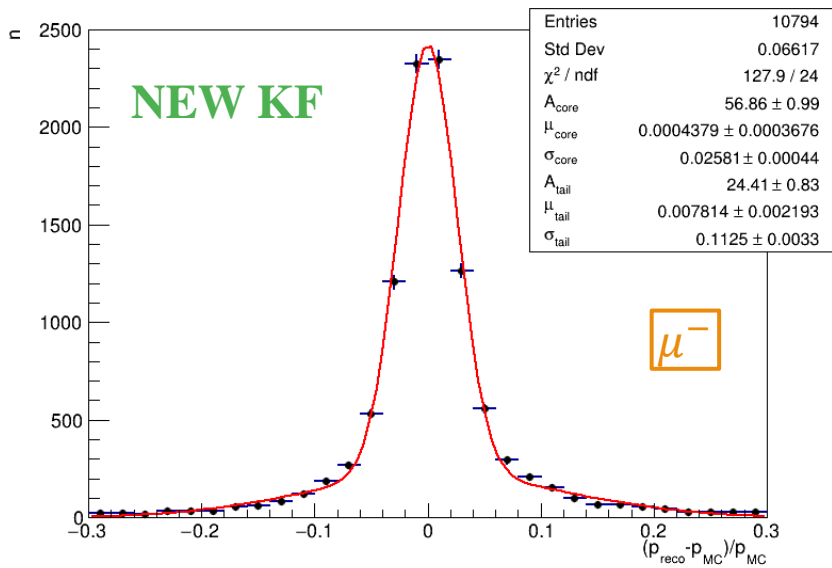
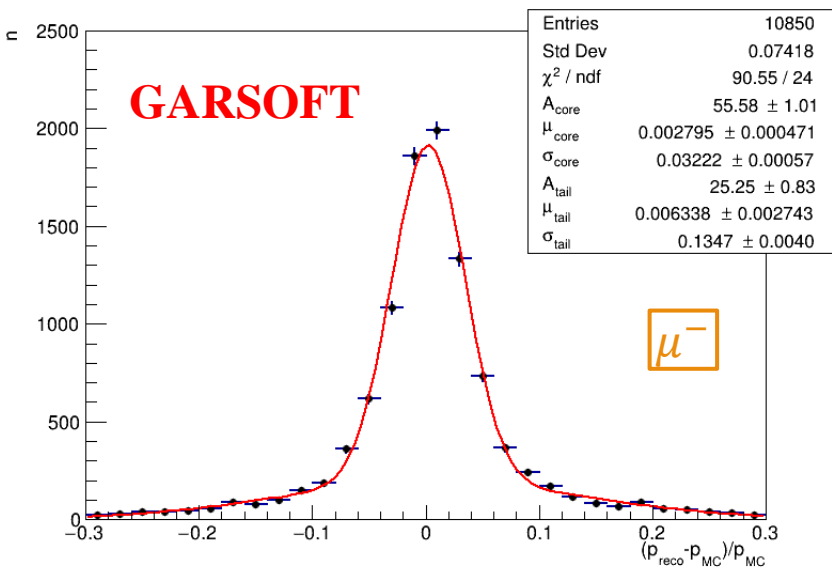
- Momentum resolution should go as  $\propto 1/\sqrt{NPoints}$  (<https://indico.fnal.gov/event/58350/> )
- Old study seemed to indicate that the new KF out-performed the current garsoft one only for long tracks: new study shows new KF outperforms old one over the whole spectrum.
- Two major bugs found:
  - In track point ordering wrong cutoff parameter was used, reducing length of tracks (fSortDistCut = 10cm instead of 20 cm)
  - Cross Length between points was calculated incorrectly for energy loss corrections

# MUONS MOMENTUM RESOLUTION : GARSOFT VS NEW KF VS CDR



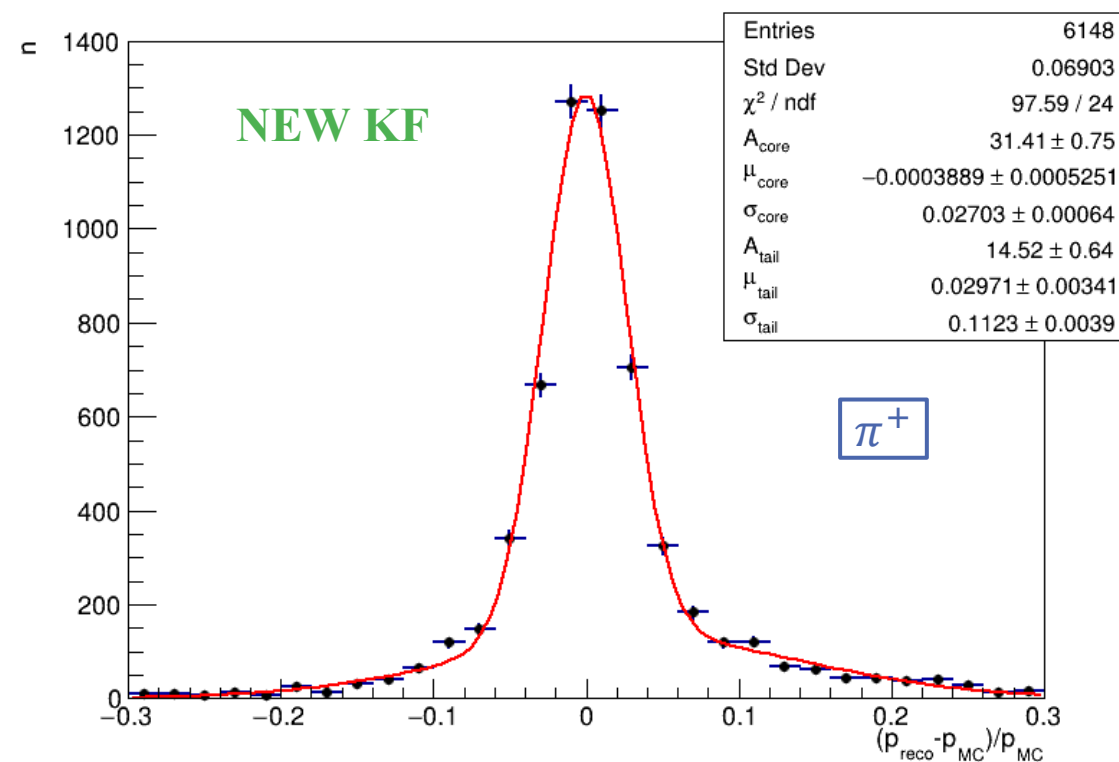
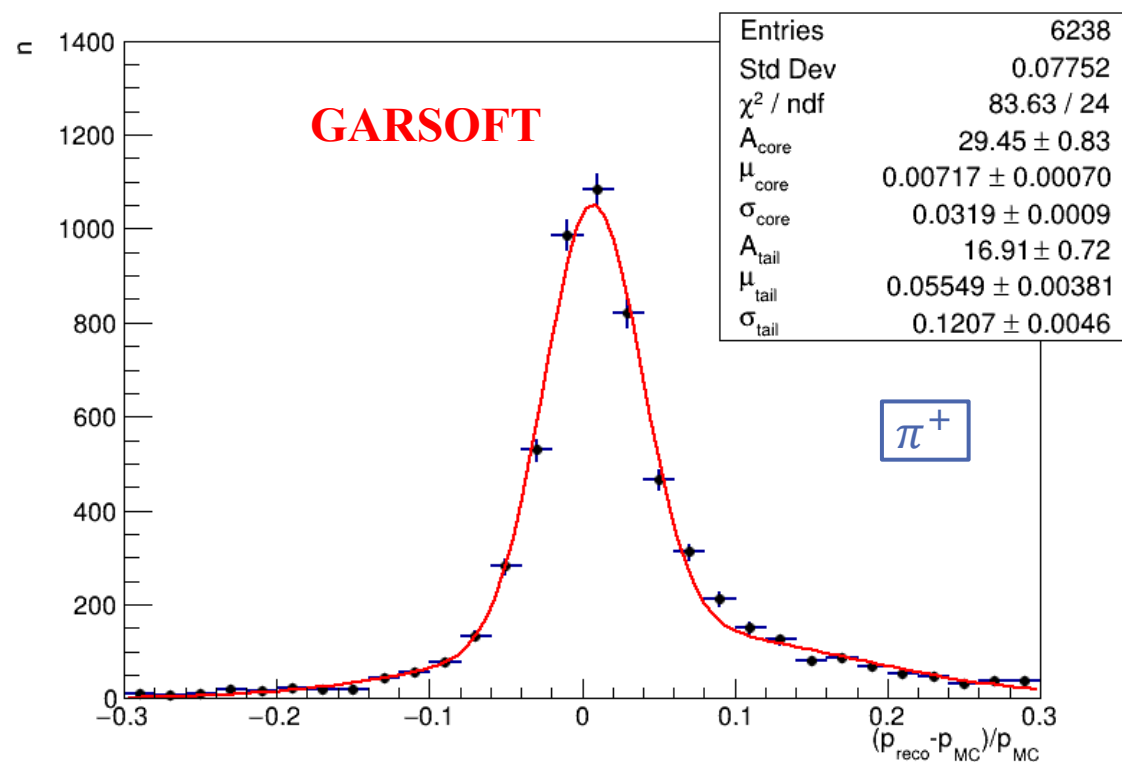
- Momentum fractional residuals for **muon sample** define reconstruction and resolution biases
- $(p_{\text{reco}} - p_{\text{MC}}) / p_{\text{MC}}$  distributions are fitted with a double Gauss fit, like in CDR, defining a core and tails sample:
  - **GArSoft**:  $(A_{\text{core}}, \mu_{\text{core}}, \sigma_{\text{core}}) = (56, 0.3\%, 3.2\%)$      $(A_{\text{tails}}, \mu_{\text{tails}}, \sigma_{\text{tails}}) = (25, 0.6\%, 13\%)$
  - **New KF**:  $(A_{\text{core}}, \mu_{\text{core}}, \sigma_{\text{core}}) = (57, 0.04\%, 2.6\%)$      $(A_{\text{tails}}, \mu_{\text{tails}}, \sigma_{\text{tails}}) = (24, 0.8\%, 11\%)$
  - **CDR**:  $(A_{\text{core}}, \mu_{\text{core}}, \sigma_{\text{core}}) = (100, -0.4\%, 3\%)$      $(A_{\text{tails}}, \mu_{\text{tails}}, \sigma_{\text{tails}}) = (49, -1.4\%, 14\%)$
- **New KF improves biases and resolutions overall for the muon sample**

# MUONS MOMENTUM RESOLUTION : GARSOFT VS NEW KF VS CDR



**NOTE:** tracks with  $N_{\text{points}} < 50$  excluded from the sample because highly problematic; Results much more in agreement with CDR compared to previous study for which the portion of core sample was closer to 1/2 than 2/3

# PIONS MOMENTUM RESOLUTION : GARSOFT VS NEW KF VS CDR

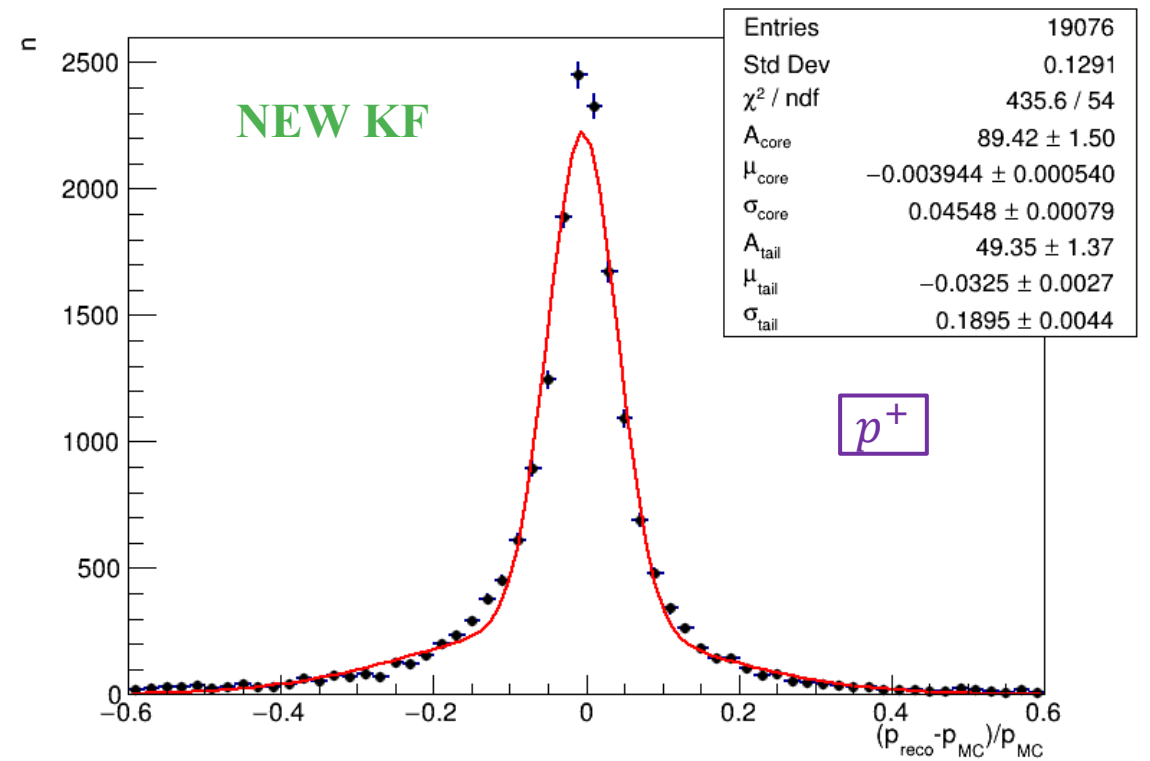
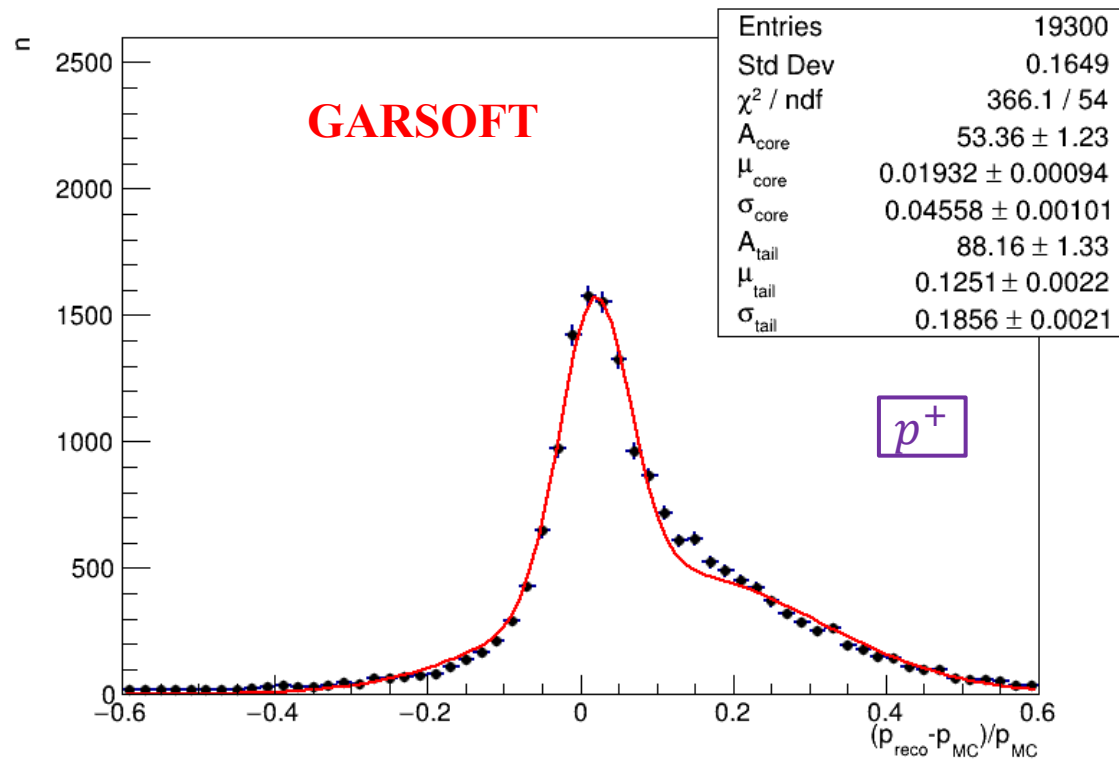


•  $(p_{\text{reco}} - p_{\text{MC}}) / p_{\text{MC}}$  distributions are fitted with a double Gauss fit, like in CDR, defining a core and tails sample:

- **GArsSoft:**  $(A_{\text{core}}, \mu_{\text{core}}, \sigma_{\text{core}}) = (30, 0.7\%, 3.2\%)$      $(A_{\text{tails}}, \mu_{\text{tails}}, \sigma_{\text{tails}}) = (17, 6\%, 12\%)$
- **New KF:**  $(A_{\text{core}}, \mu_{\text{core}}, \sigma_{\text{core}}) = (31, -0.04\%, 2.7\%)$      $(A_{\text{tails}}, \mu_{\text{tails}}, \sigma_{\text{tails}}) = (15, 3\%, 11\%)$

- **New KF improves biases and resolutions overall for the pion sample very similarly to muon sample (similar dEdx and mass)**
- **NOTE1:** No CDR Results available: no direct comparison possible
- **NOTE2:** Tracks with less than 50 points are removed as for the muons

# PROTONS MOMENTUM RESOLUTION : GARSOFT VS NEW KF



•  $(p_{\text{reco}} - p_{\text{MC}}) / p_{\text{MC}}$  distributions are fitted with a double Gauss fit, like in CDR, defining a core and tails sample:

- **GArSoft:**  $(A_{\text{core}}, \mu_{\text{core}}, \sigma_{\text{core}}) = (53, 2\%, 4.6\%)$      $(A_{\text{tails}}, \mu_{\text{tails}}, \sigma_{\text{tails}}) = (88, 12\%, 19\%)$
- **New KF:**  $(A_{\text{core}}, \mu_{\text{core}}, \sigma_{\text{core}}) = (89, -0.4\%, 4.5\%)$      $(A_{\text{tails}}, \mu_{\text{tails}}, \sigma_{\text{tails}}) = (49, -3\%, 19\%)$

- **New KF majorly improves biases and resolutions for the proton sample (similar dEdx and mass)**
- **NOTE1:** No CDR Results available: no direct comparison possible
- **NOTE2:** Tracks with less than 50 points are removed as for the muons



# RESOLUTION DEPENDENCIES

- Analytical formulas derived directly from PDG chapter 34 on detectors give realistic expectations for the  $q/p_T$  resolution dependency: <https://pdg.lbl.gov/2019/reviews/rpp2019-rev-particle-detectors-accel.pdf>

## POINT RESOLUTION

$$\sigma_N(q/p_T) = \frac{q\sigma_r}{0.3BL_{arm}} \sqrt{\frac{720}{N}}$$

## MULTIPLE SCATTERING

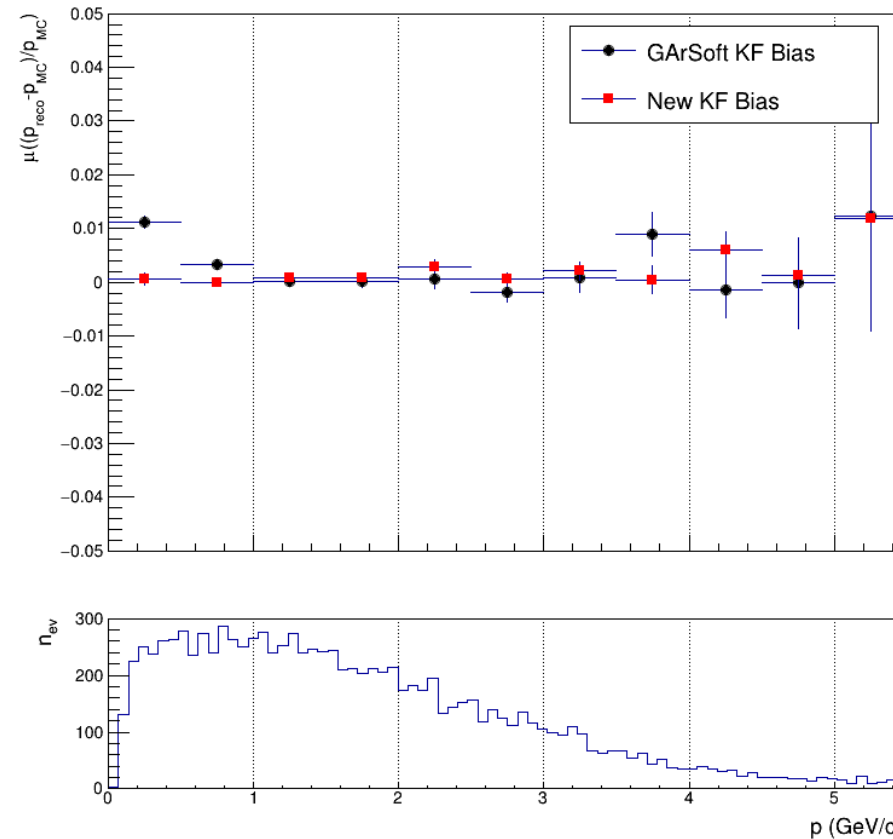
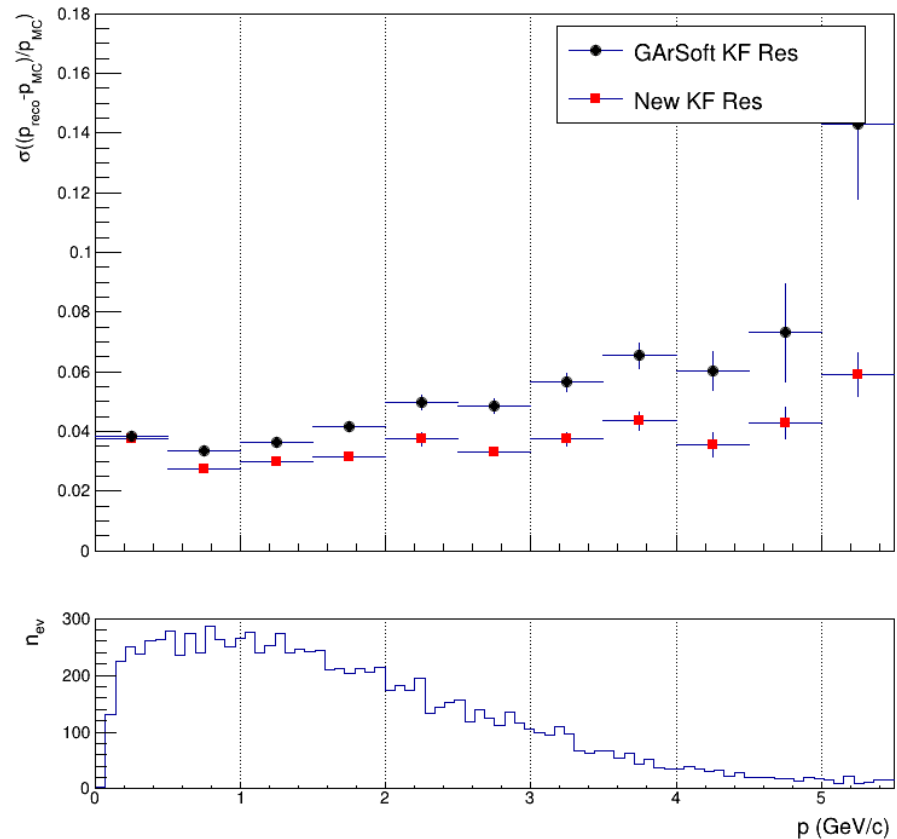
$$\sigma_{MS}(q/p_T) = \left\langle \frac{1}{\beta p_T} \right\rangle \frac{q \times 0.016 \times B \times 0.3}{L} \sqrt{\frac{L}{X_0}}$$

$$\sigma_{ana}(q/p_T) = \sqrt{\sigma_N^2 + \sigma_{MS}^2}$$

- $\sigma_r$  = radial resolution
- $B$  = magnetic field
- $N$  = number of points measured
- $L_{arm}$  = Lever arm on XY plane
- $L$  = Length of the track on XY plane
- $X_0$  = Radiation length in cm
- $\beta$  = velocity
- NOTE:**  $\langle 1/\beta p_T \rangle$  = value of  $1/(\beta p_T)$  averaged along the trajectory to take into account energy loss

NB:  $q/p_T$  scaling for high density materials, such as ND-GAr's gas mixture, should be dominated by the  $\sigma_{MS}$  component

# MOMENTUM RESOLUTION AND BIAS VS P: MUONS $\mu^-$

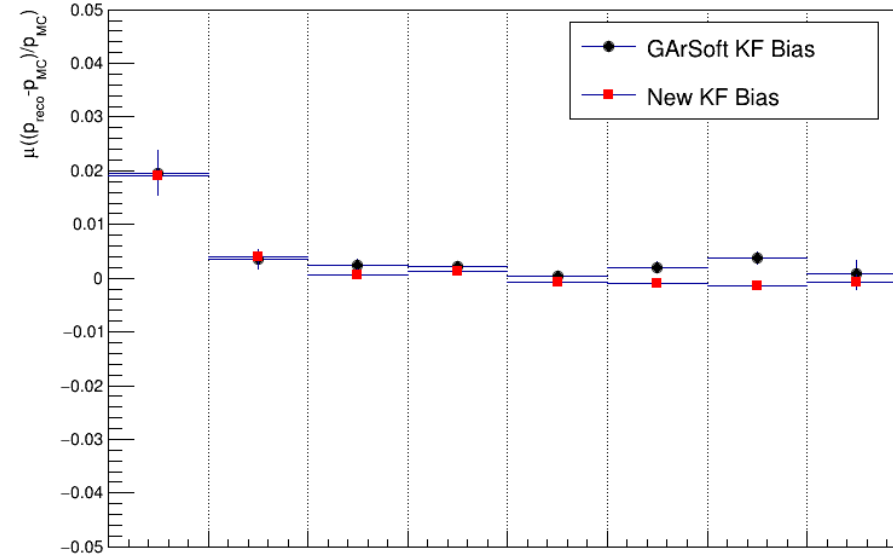
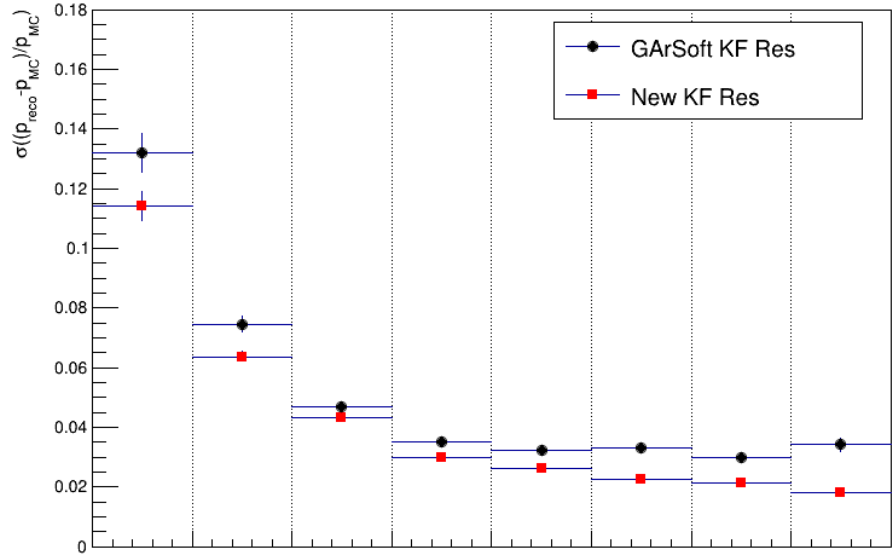


Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice

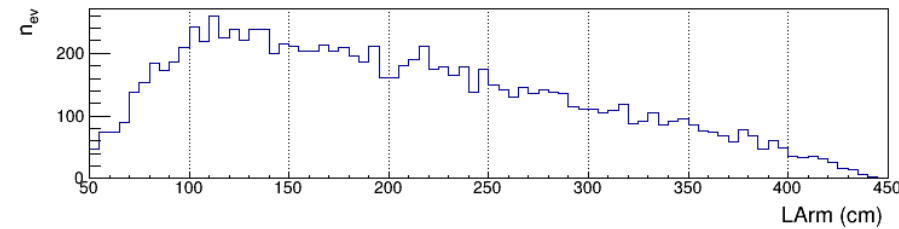
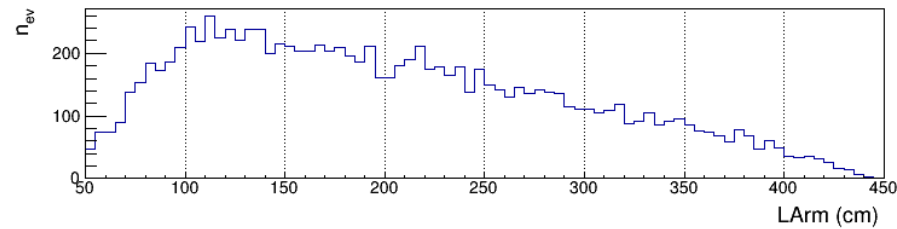
Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{\text{points}} < 50$  are cut)

- Momentum resolution should be mostly momentum independent in this range and at these densities. This is largely true for the new KF but not in garsoft
- Note that the  $p_T$  should be averaged through the whole track, which wasn't done here.
- New KF improves resolution over the whole spectrum and keeps bias mostly the same

# MOMENTUM RESOLUTION AND BIAS VS LARM: MUONS $\mu^-$



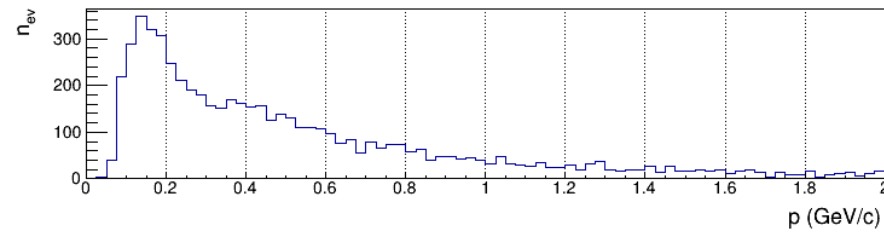
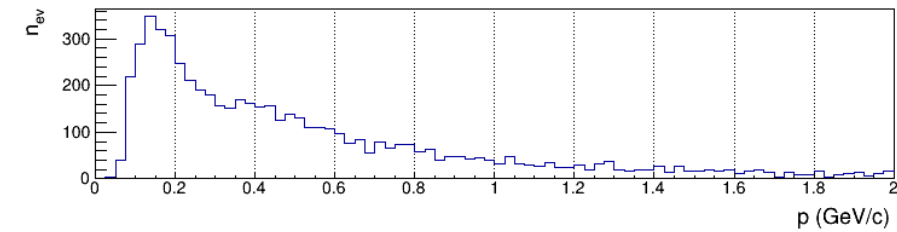
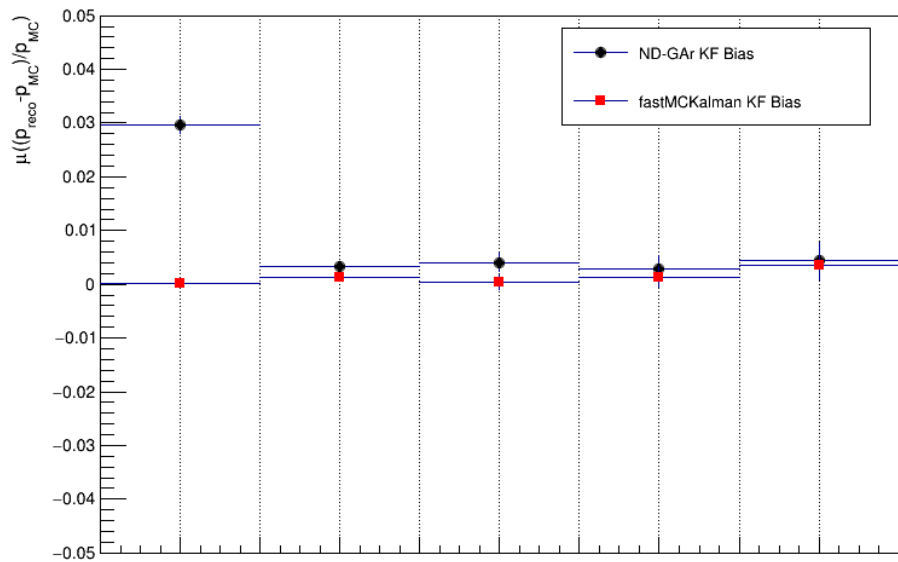
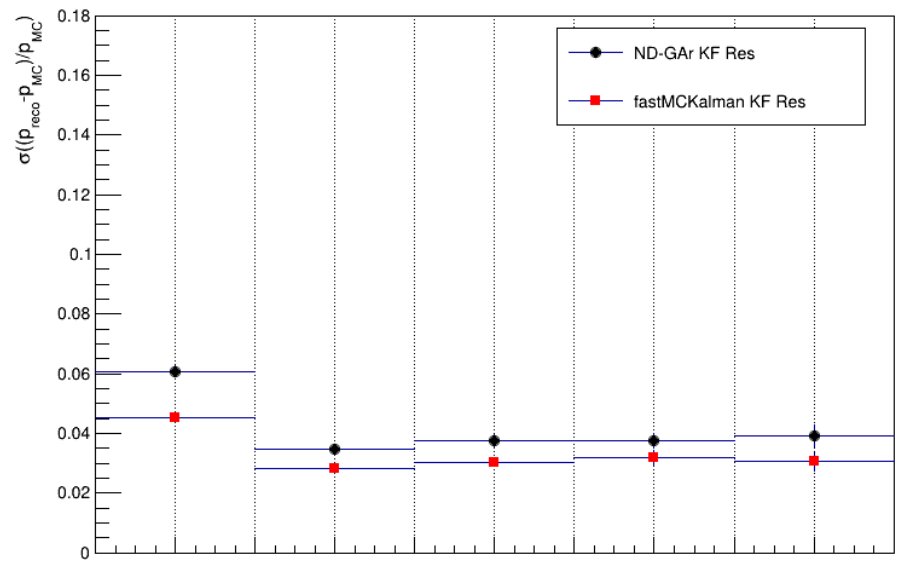
Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each p slice



Correspondent p (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

- Lever Arm: distance in transverse (yz) plane between first and last point in the track
- Momentum resolution in the range  $p \in [0,6]$  GeV/c should be multiple scattering dominated and go as  $\propto 1/\sqrt{LArm}$  (dependencies on Npoints and Length are similar; see back-up)
- New KF improves resolution over the whole spectrum and keeps bias mostly the same

# MOMENTUM RESOLUTION AND BIAS VS P: PIONS $\pi^+$

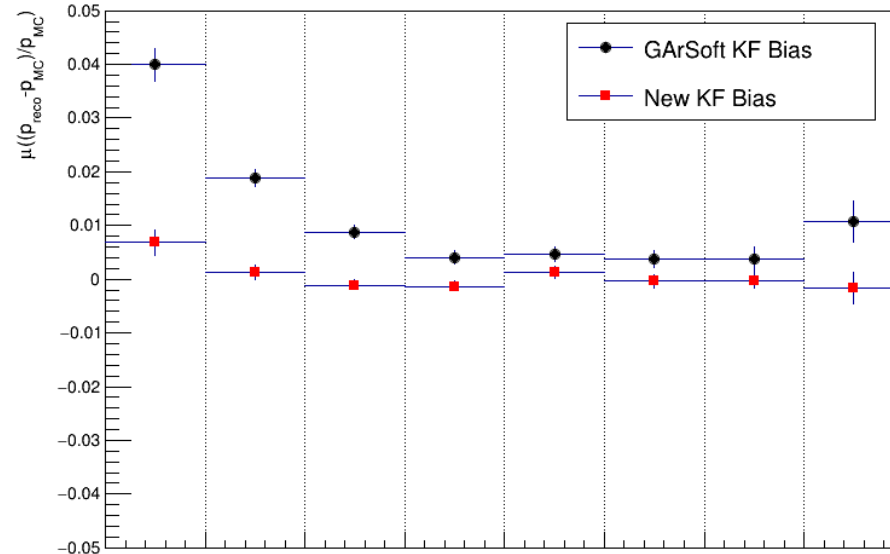
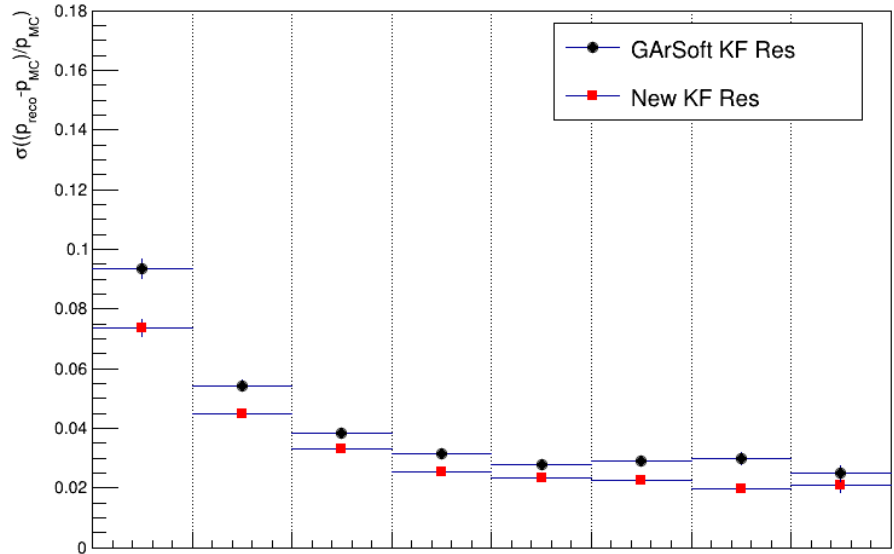


Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each p slice

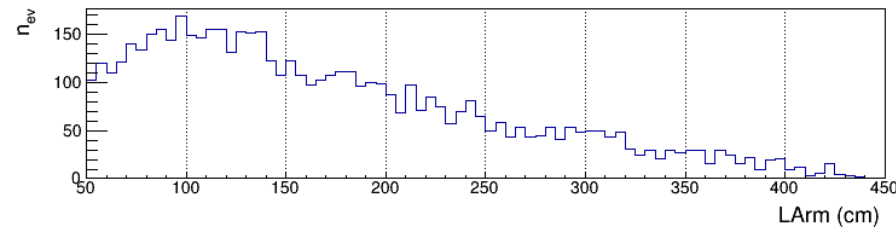
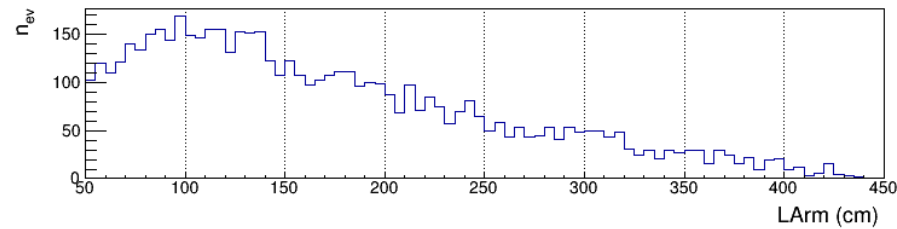
Correspondent p (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

- Momentum resolution should be mostly momentum independent in this range and at these densities.
- Primary pions are on average much lower in momentum than muons
- NB: at lower momenta, for higher mass particles the tracks will tend to be shorter and the resolution will degrade
- NB: pT should be averaged through the whole track, which wasn't done here.
- New KF improves resolution over the whole spectrum as well as the bias

# MOMENTUM RESOLUTION AND BIAS VS LARM: PIONS



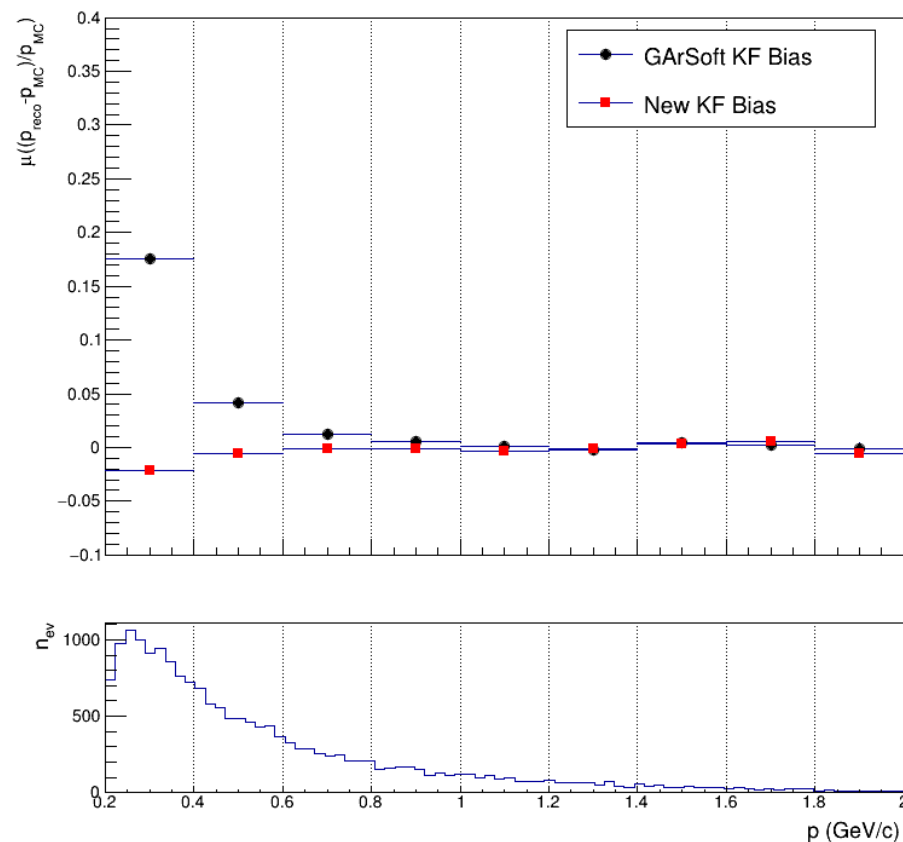
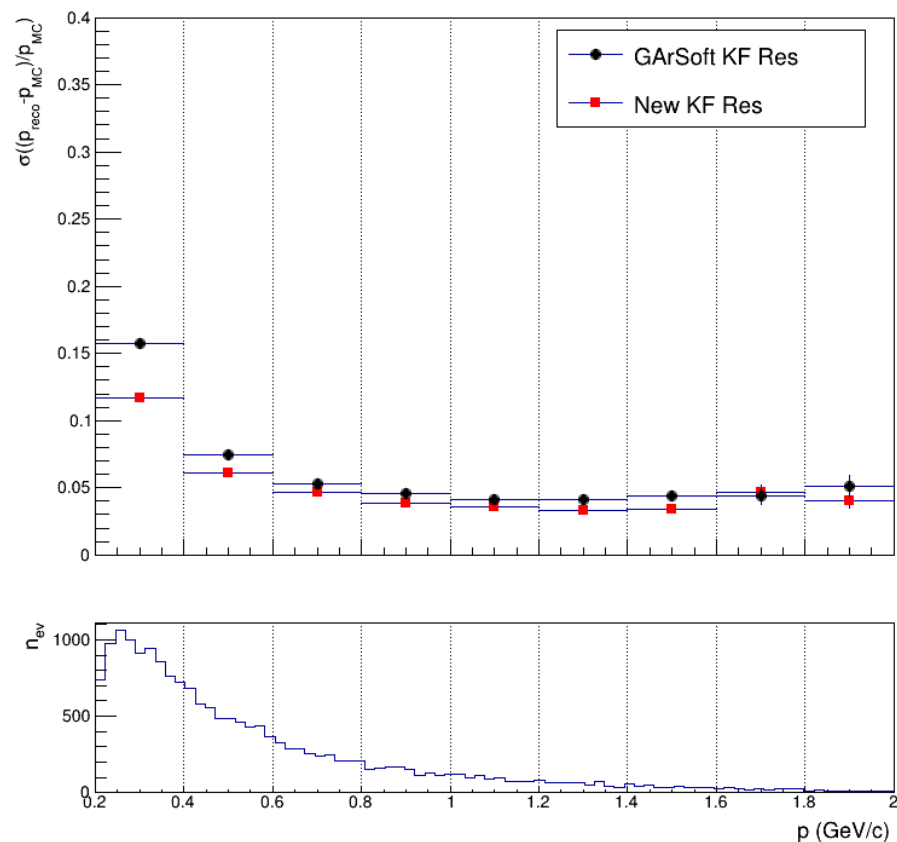
Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice



Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

- Lever Arm: distance in transverse ( $yz$ ) plane between first and last point in the track
- Momentum resolution in the range  $p \in [0,6]$  GeV/c should be multiple scattering dominated and go as  $\propto 1/\sqrt{LArm}$
- New KF improves resolution over the whole spectrum as well as the bias: as the mass of the pions is higher than the one of the muons, the dEdx and MS components of the new KF start having a bigger effect

# MOMENTUM RESOLUTION AND BIAS VS P: PROTONS $p^+$



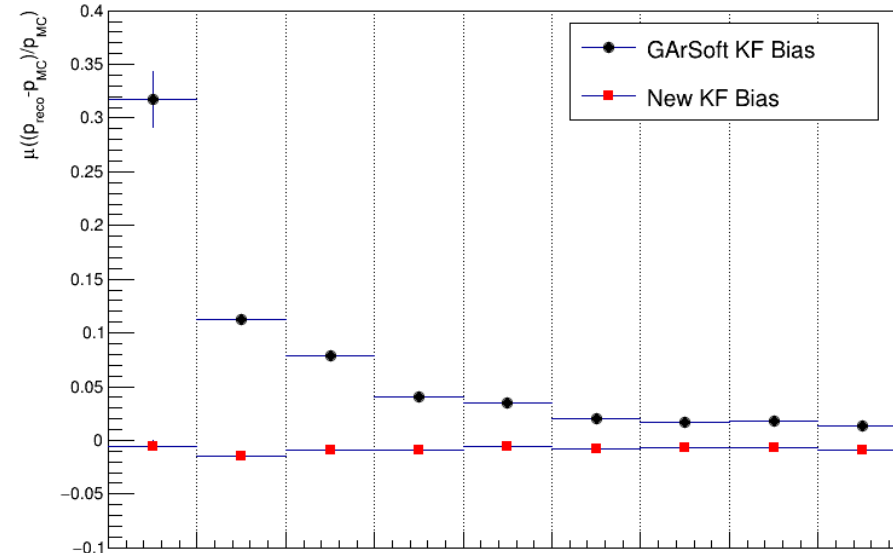
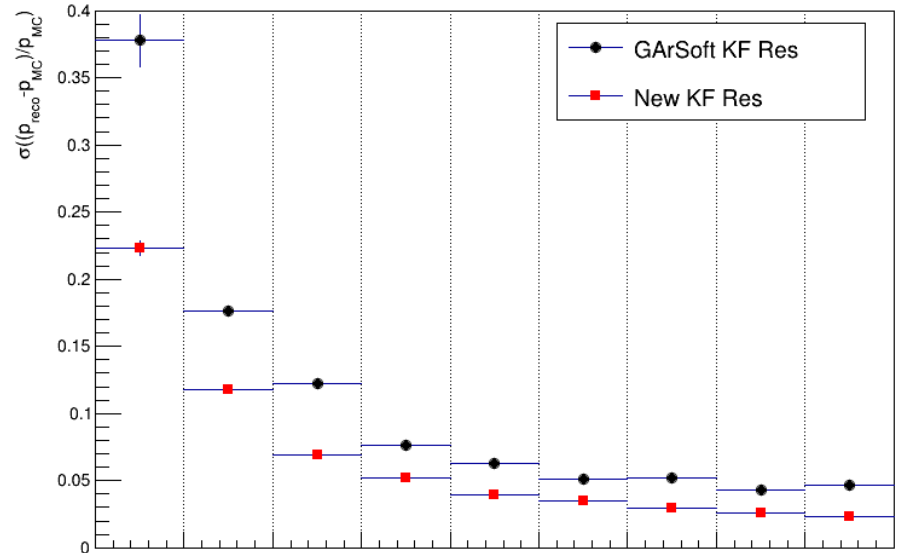
Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice

NB: Y axis range is wider as the resolution is worse for protons

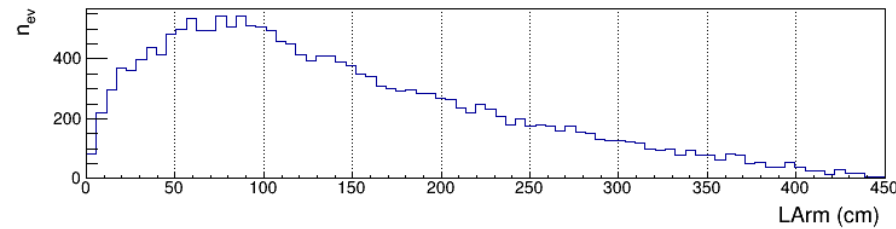
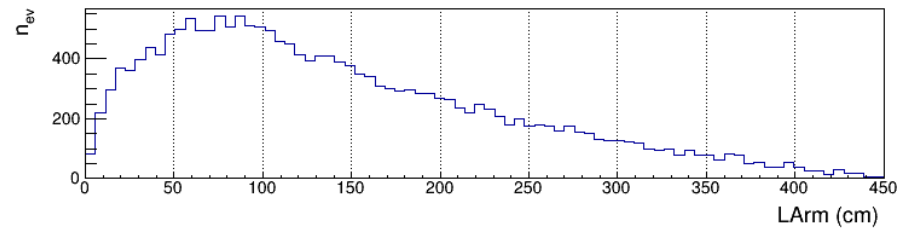
Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{\text{points}} < 50$  are cut)

- Momentum resolution should be mostly momentum independent in this range and at these densities.
- Primary protons are on average much lower in momentum than muons and much more similar to pions
- NB: at lower momenta, for higher mass particles the tracks will tend to be shorter and the resolution will degrade
- NB:  $p_T$  should be averaged through the whole track, which wasn't done here.
- New KF improves resolution over the whole spectrum and especially the bias

# MOMENTUM RESOLUTION AND BIAS VS LARM: PROTONS $p^+$



Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice



Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

- Lever Arm: distance in transverse ( $yz$ ) plane between first and last point in the track
- Momentum resolution in the range  $p \in [0,6]$  GeV/c should be multiple scattering dominated and go as  $\propto 1/\sqrt{LArm}$
- New KF improves resolution over the whole spectrum as well as the bias: as the mass of the protons is higher than the one of the muons, the dEdx and MS components of the new KF start having a bigger effect

# SUMMARY AND CONCLUSIONS

- New ALICE-BASED Kalman Filter was tested and compared to the current GArSoft Reconstruction over a sample of primary particles from  $\nu_\mu CC$  interactions with MC vertex in TPC fiducial volume:
  - Selected from a sample of  $4.35 \times 10^4$  neutrino interactions in active TPC volume;
  - Produced using GENIE module in GArSoft v2\_18\_00 with standard flux;
  - Primary protons, pions and muons were considered
- Main Takeaways:
  1. After bug fixes, new KF shown to provide significant performance benefits for all analyzed particle types from the core sample of  $\nu_\mu CC$  interactions
  2. Proton reconstruction is especially biased at the current state and the new KF can improve this
- Next steps:
  1. Finish improving pull tests so that they are as expected for all particle types (not discussed in this presentation)
  2. Explore benefits of the improved performance (e.g. TKI hydrogen study <https://indico.fnal.gov/event/59667/> )
  3. Implement in GArSoft



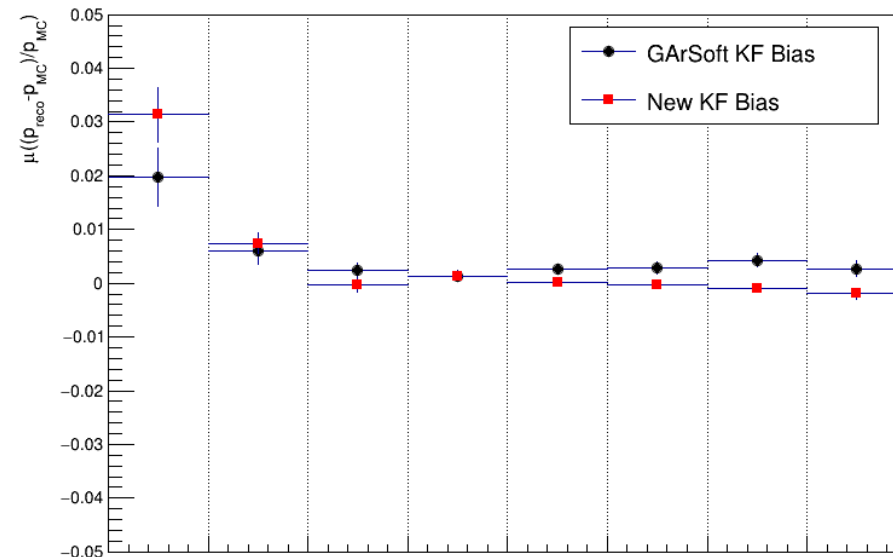
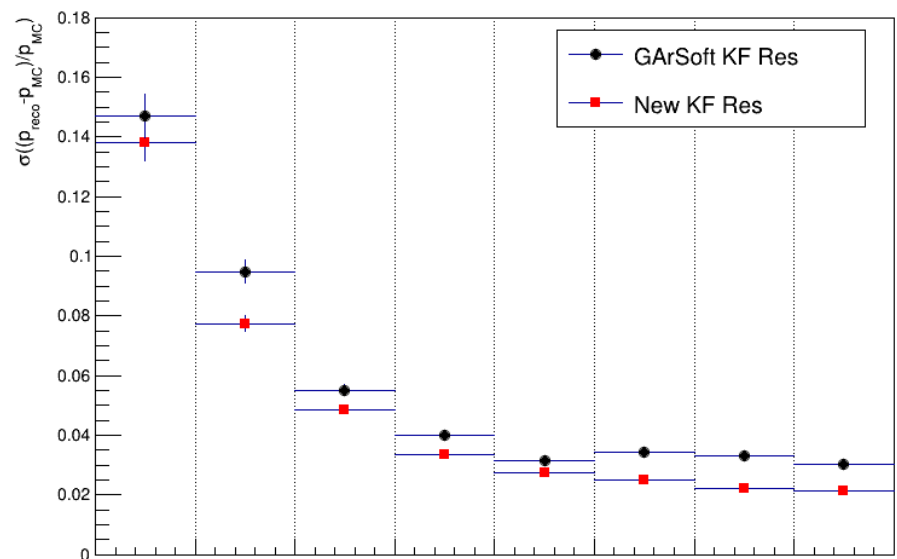


**THANK YOU**

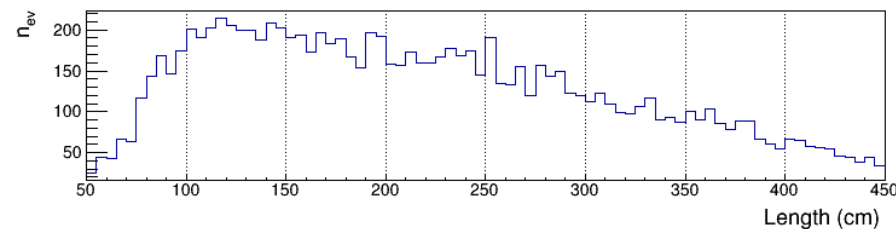
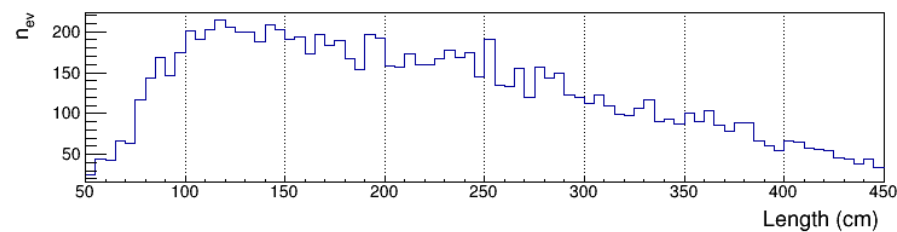
ROBERT RATHBUN WILSON

# BACK-UP

# MOMENTUM RESOLUTION AND BIAS VS LENGTH: MUONS $\mu^-$



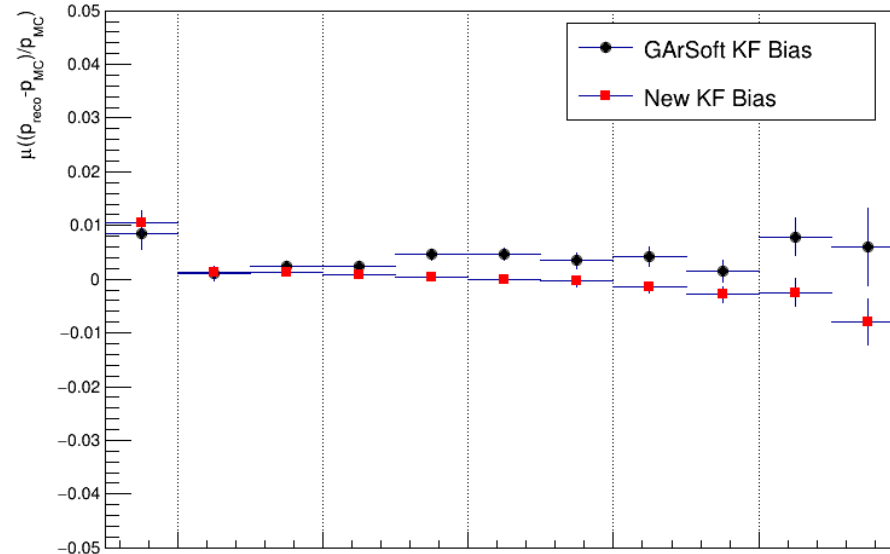
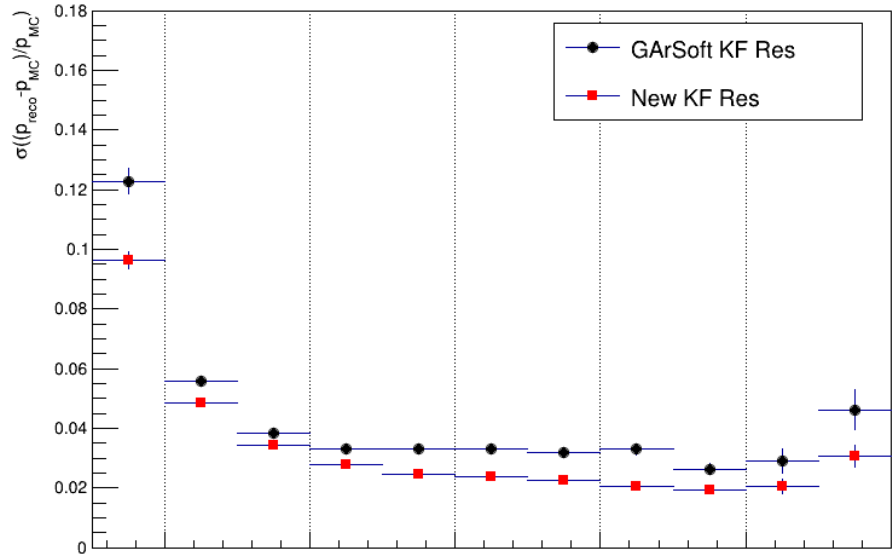
Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice



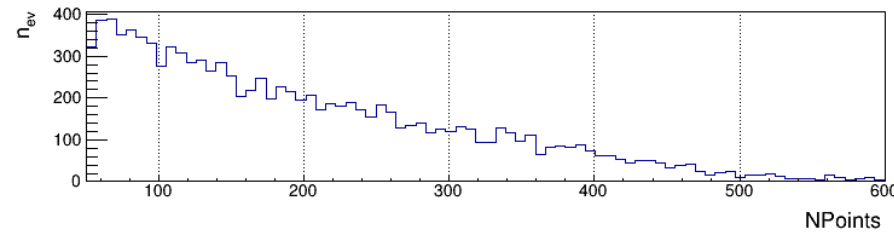
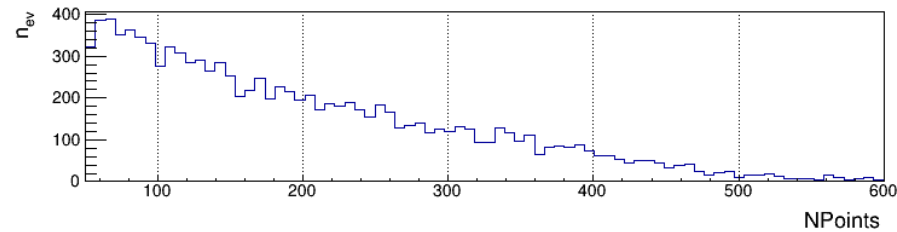
Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

- Momentum resolution in the range  $p \in [0,6]$  GeV/c should be multiple scattering dominated and go as  $\propto 1/\sqrt{Length}$
- New KF improves resolution over the whole spectrum and keeps bias mostly the same

# MOMENTUM RESOLUTION AND BIAS VS NPOINTS: MUONS $\mu^-$



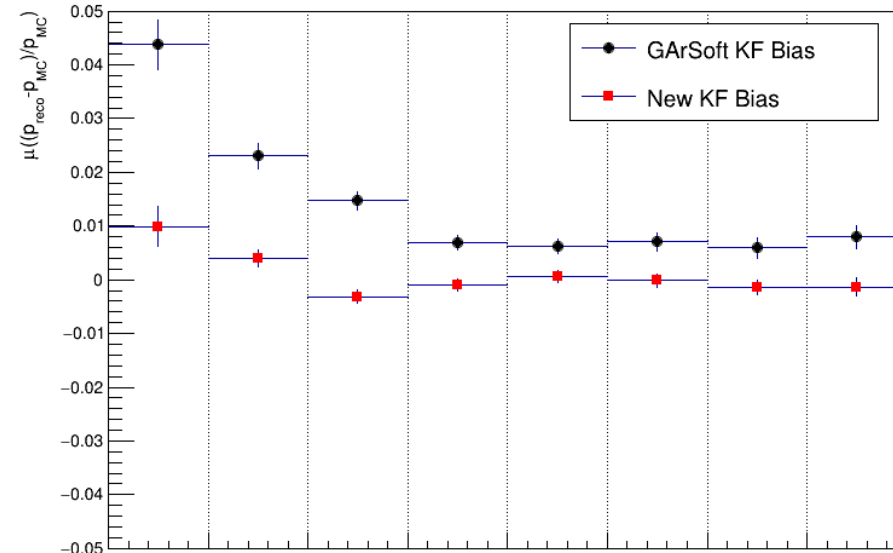
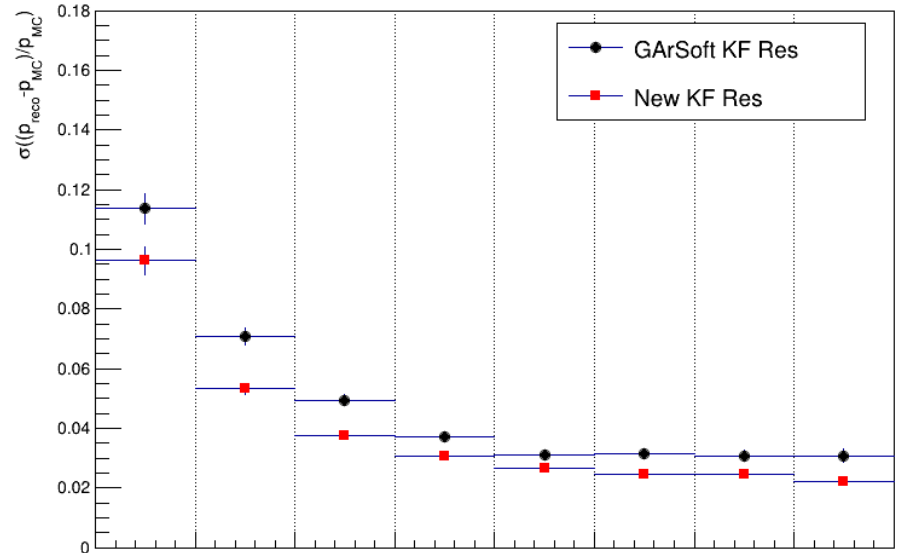
Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice



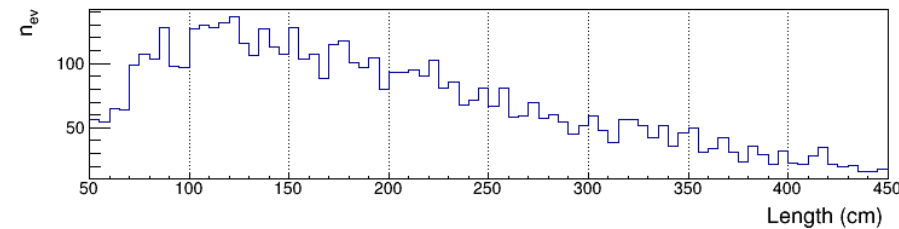
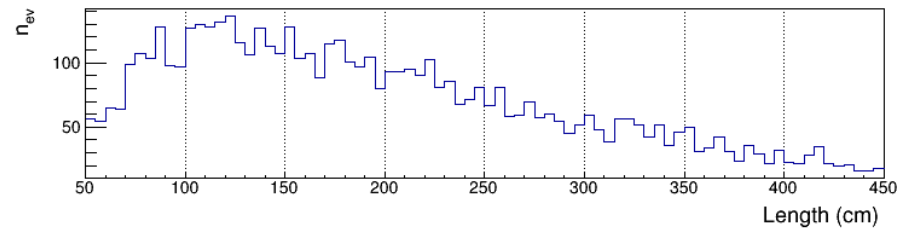
Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

- Momentum resolution in the range  $p \in [0,6]$  GeV/c should be multiple scattering dominated and go as  $\propto 1/\sqrt{NPoints}$
- New KF improves resolution over the whole spectrum and keeps bias mostly the same

# MOMENTUM RESOLUTION AND BIAS VS LENGTH: PIONS $\pi^+$



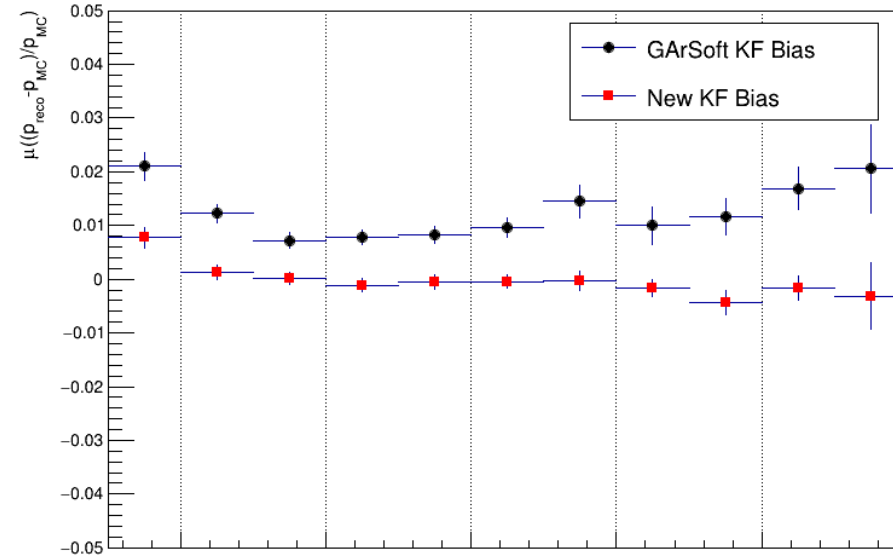
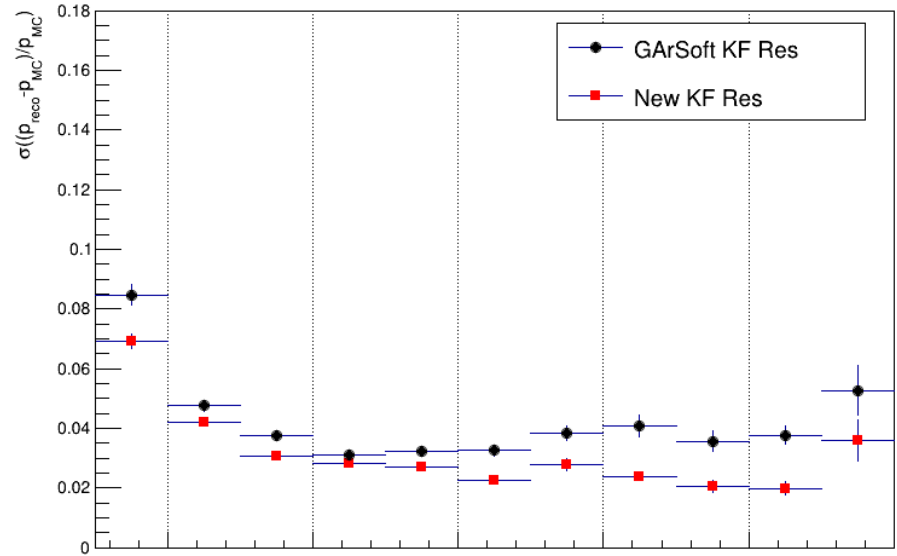
Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice



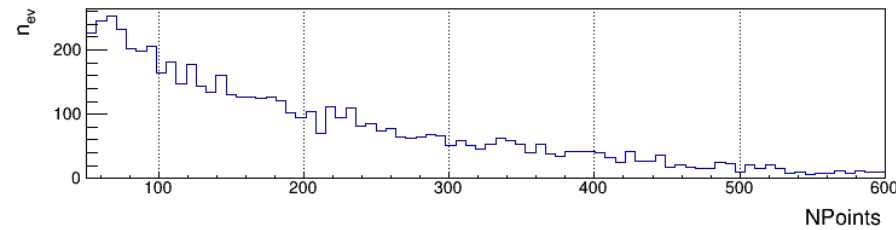
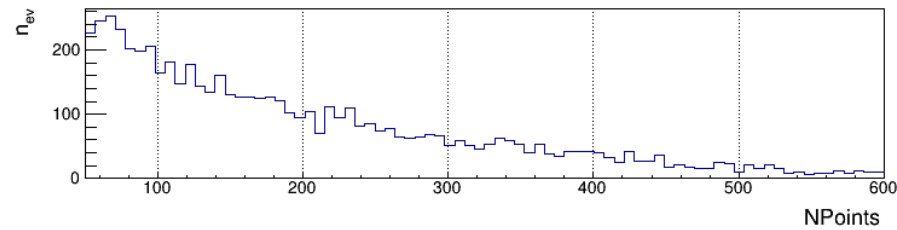
Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

- Momentum resolution in the range  $p \in [0,6]$  GeV/c should be multiple scattering dominated and go as  $\propto 1/\sqrt{Length}$
- New KF improves resolution over the whole spectrum as well as the bias: as the mass of the pions is higher than the one of the muons, the dEdx and MS components of the new KF start having a bigger effect

# MOMENTUM RESOLUTION AND BIAS VS NPOINTS: PIONS $\pi^+$



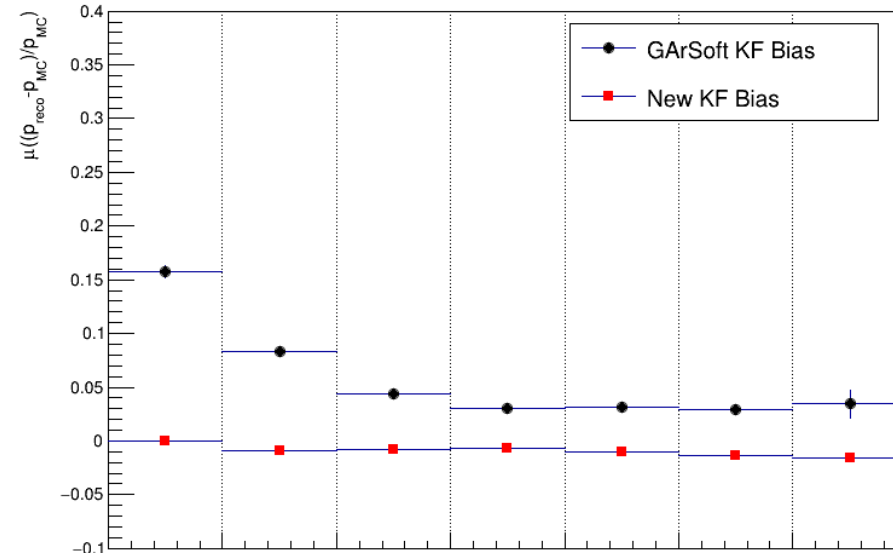
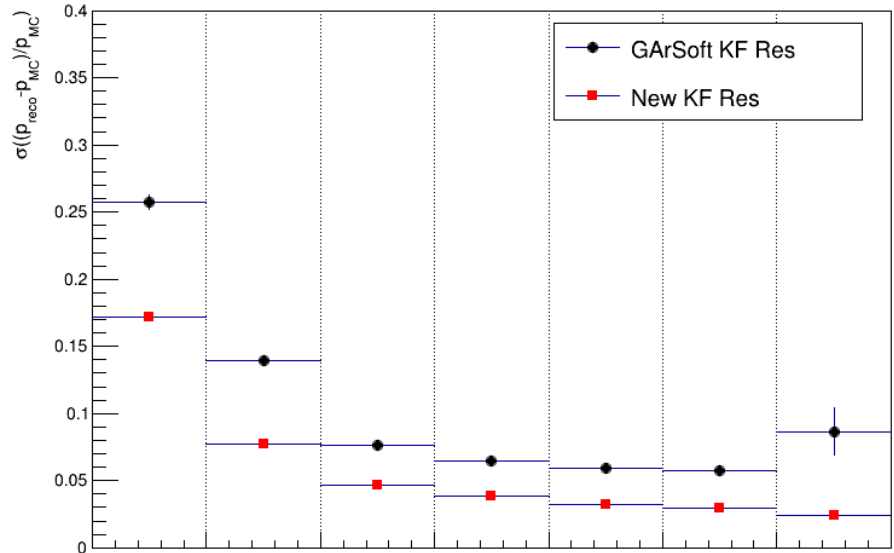
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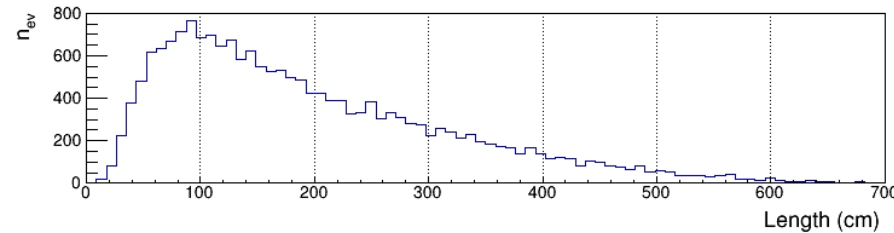
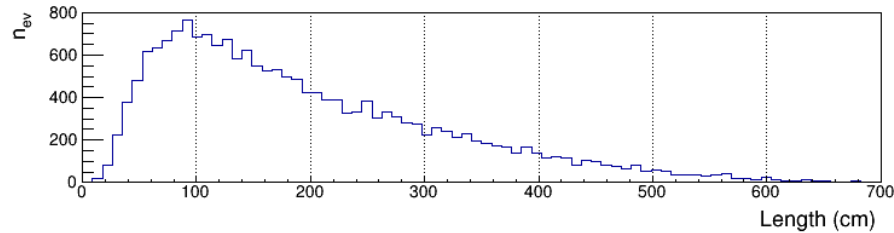
Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

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# MOMENTUM RESOLUTION AND BIAS VS LENGTH: PROTONS $p^+$



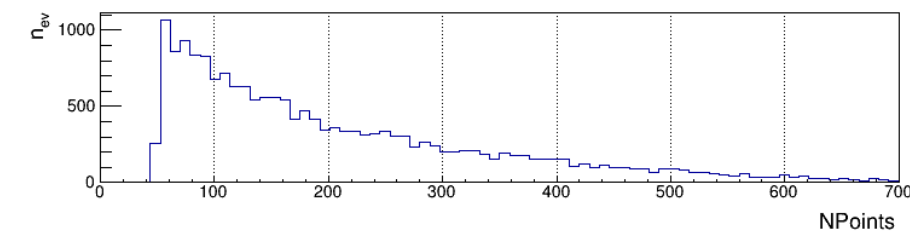
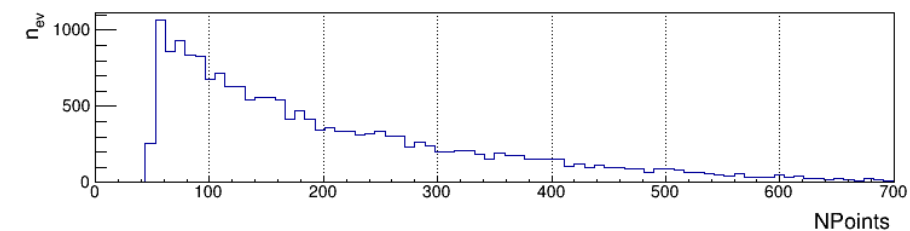
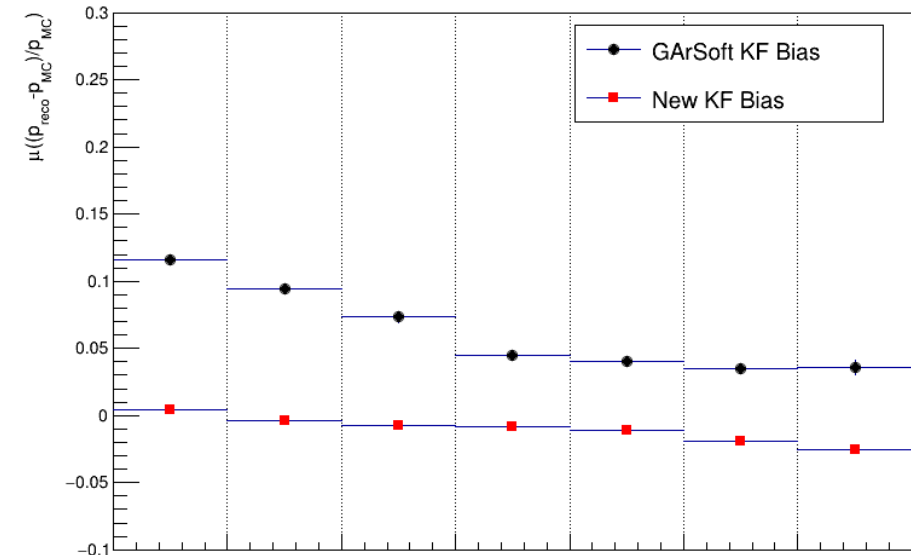
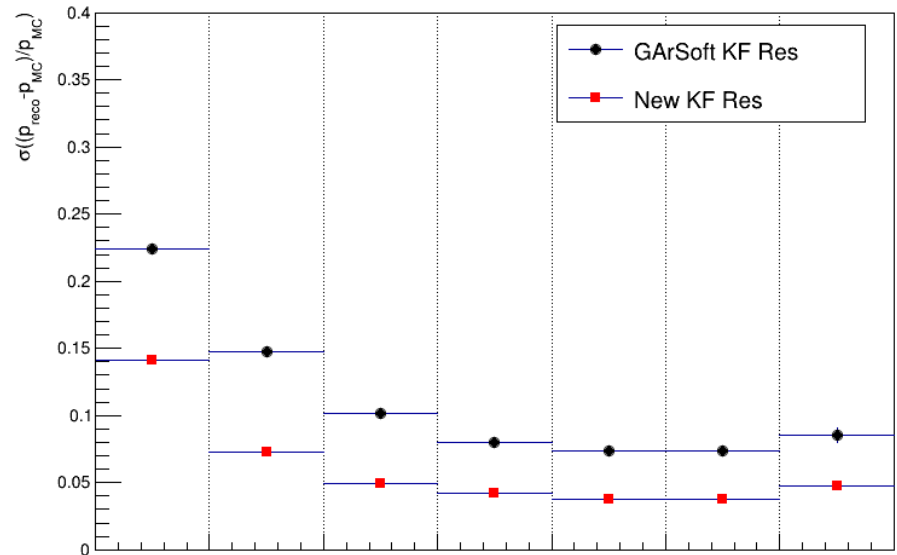
Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice



Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

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# MOMENTUM RESOLUTION AND BIAS VS NPOINTS: PROTONS $p^+$



Profile plots for resolution and bias :  $(\sigma, \mu)$  from momentum residual Gauss fit in each  $p$  slice

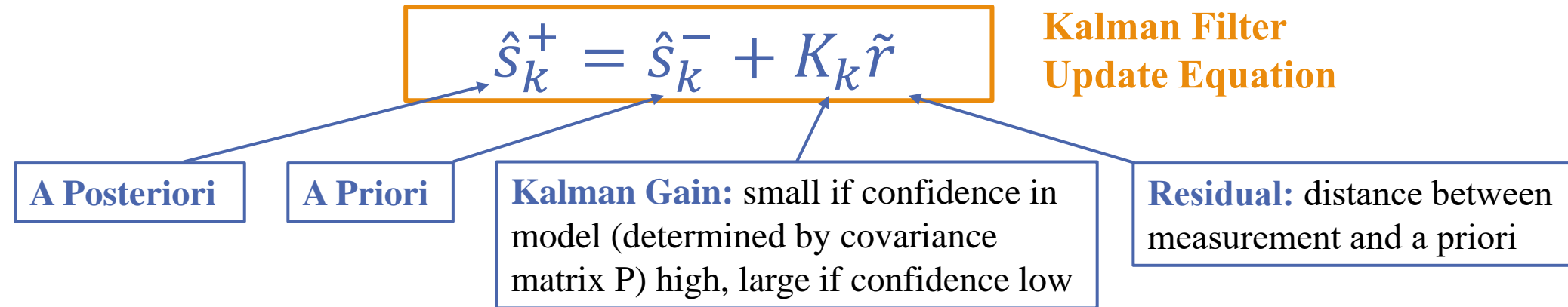
Correspondent  $p$  (GeV/c) distribution (NB: Tracks with  $N_{points} < 50$  are cut)

- Momentum resolution in the range  $p \in [0,6]$  GeV/c should be multiple scattering dominated and go as  $\propto 1/\sqrt{NPoints}$
- New KF improves resolution over the whole spectrum as well as the bias



# RECONSTRUCTION

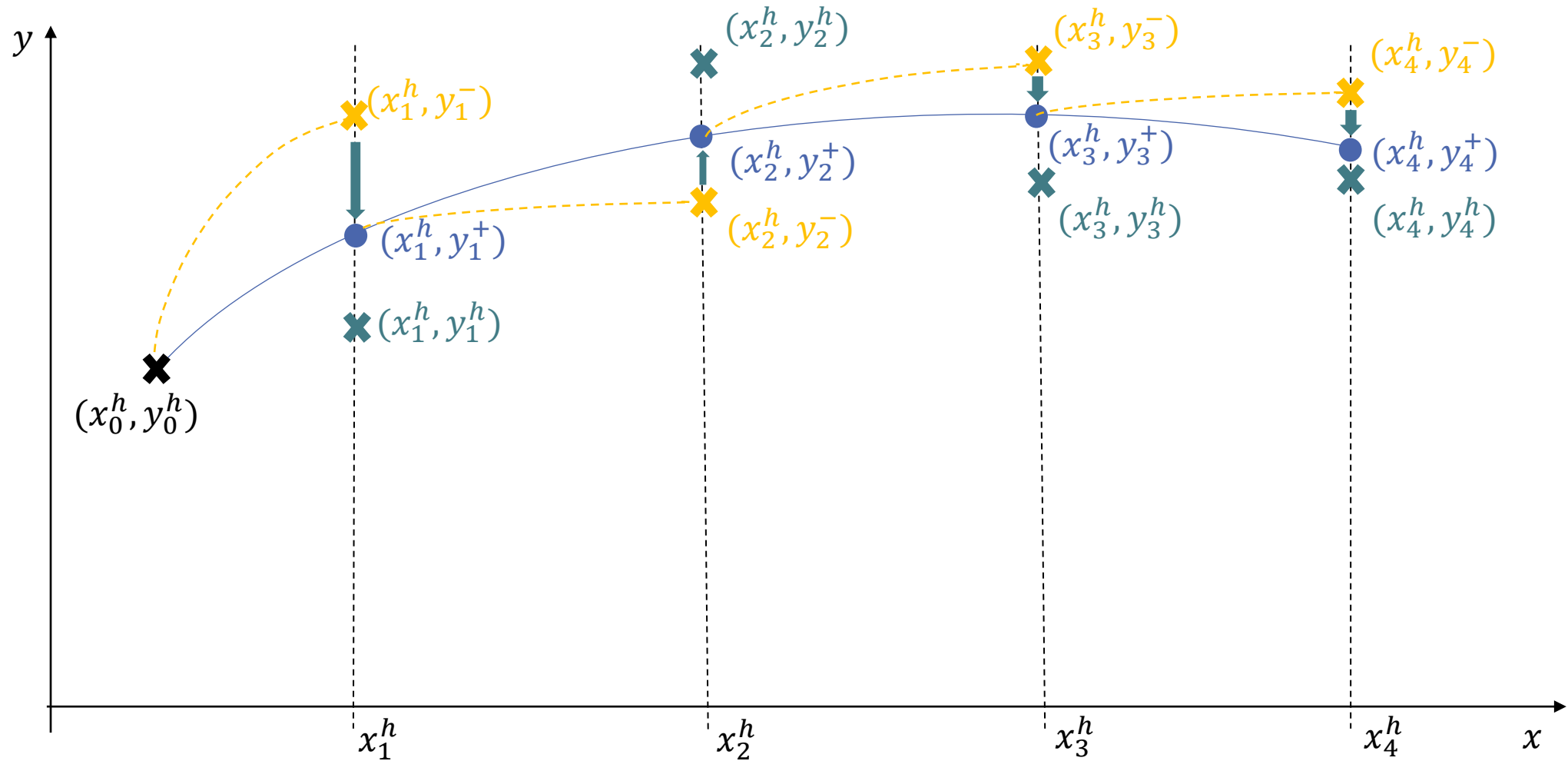
# KALMAN FILTER BASICS



- **Kalman filter:** iterative Bayesian algorithm which mediates between system knowledge and measurement. Each iteration divided in three steps:
  1. Make **A Priori prediction** of the state of the system using evolution model for the particle's trajectory
  2. Calculate **Residual:** distance between measurement and prediction
  3. Mediate between the a priori prediction and the measurement calculating **Kalman Gain** and produce **A Posteriori estimate**

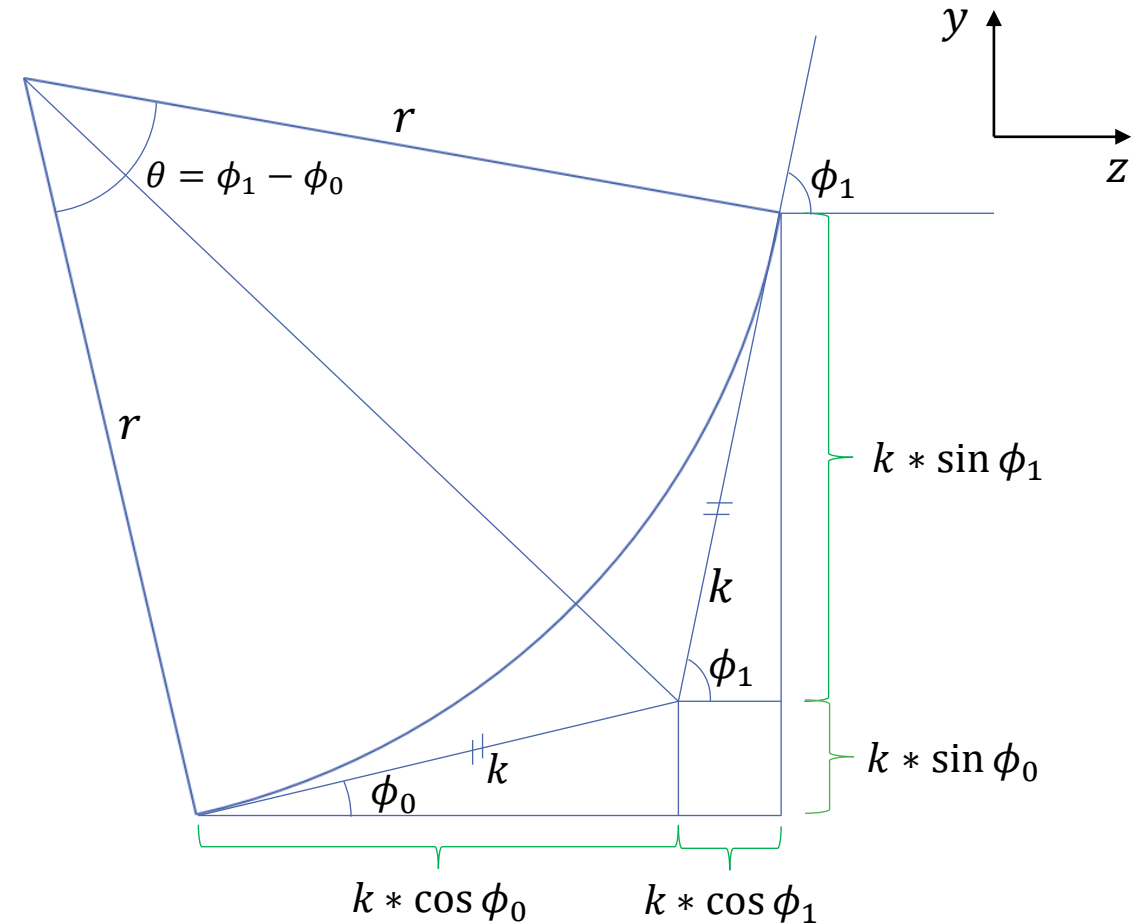
Note: See back-up for further reading

# KALMAN FILTER BASICS



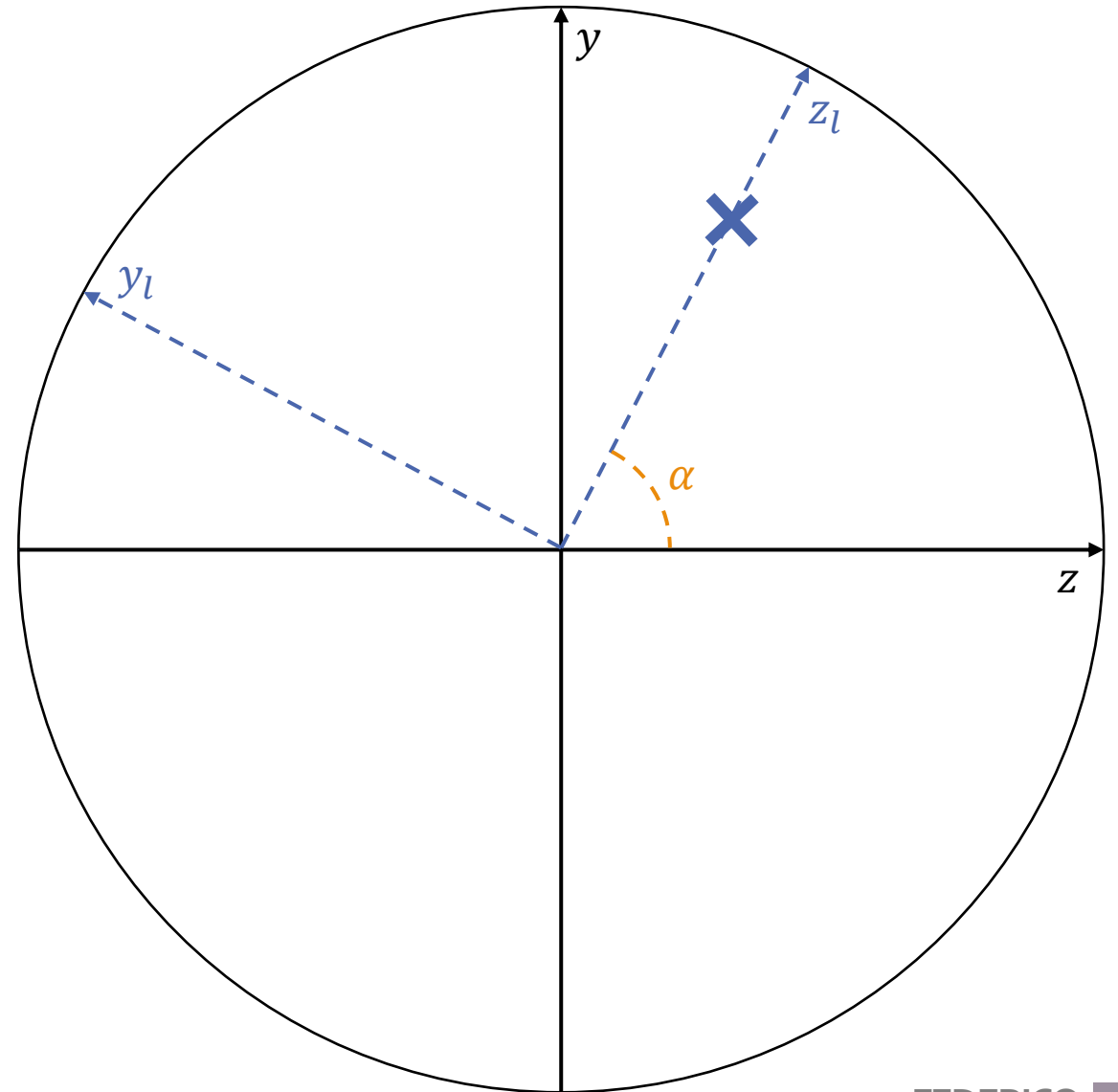
# KALMAN FILTER MODEL AND APPLICATION

- Use parametrization used in ALICE: state vector updated by the Kalman filter is  $s = (y, x, \sin\phi, \tan\lambda, \frac{q}{p_T})$
- ALICE uses no approximations in the propagation, unlike current **ND-GAr model which uses small angle approximation** (for full description check back-up and first ND-GAr-Lite presentation <https://indico.fnal.gov/event/50215/contributions/232480/> )



# KALMAN FILTER MODEL AND APPLICATION

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- Kalman filter **propagated radially**: before each propagation, the coordinate system is **rotated by an angle  $\alpha = \tan(y/z)$** , so that the track point “sits” on the local z axis (i.e. z coordinate becomes the radius from center of the detector)



# KALMAN FILTER MODEL AND APPLICATION

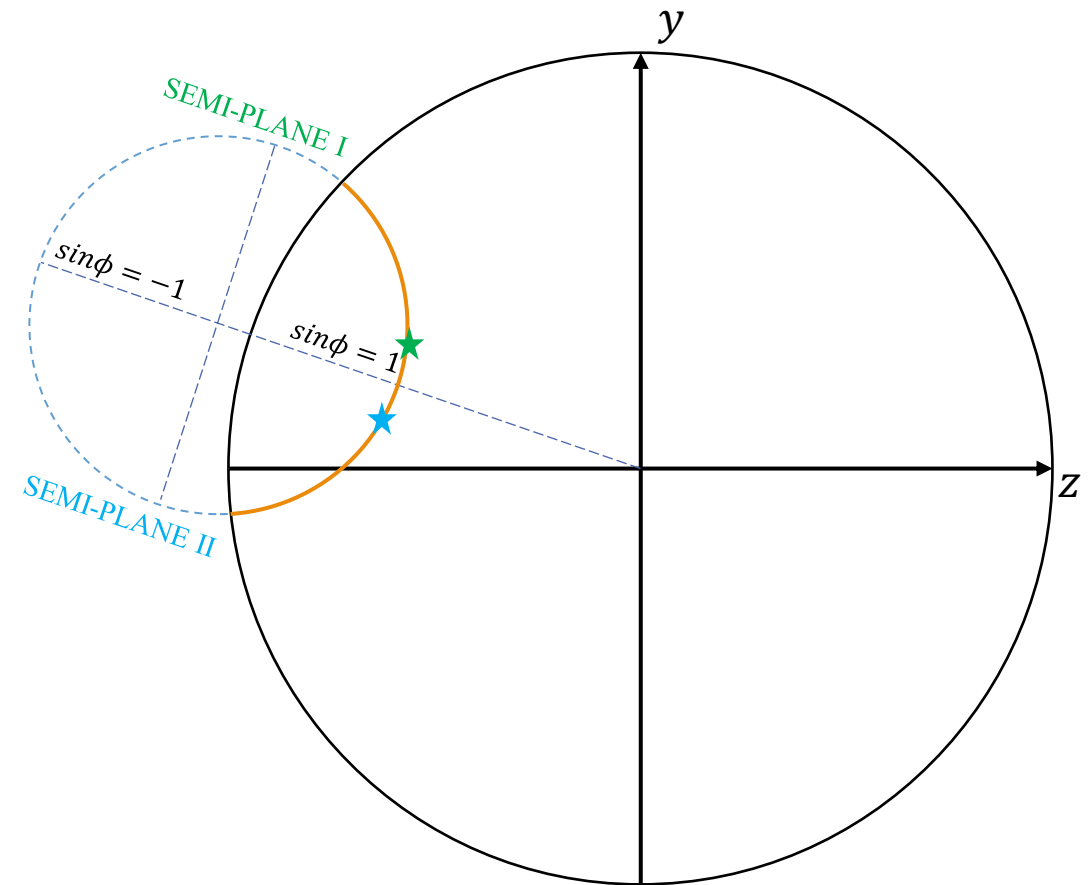
- Local  $\sin\phi$  defines two  $yz$  semi-planes with “mirrored representations”: the line separating the two is the one connecting the center of the detector and the center of curvature of the track
- As the track approaches one of the two semi-planes,  $\sin\phi$  reaches a point where it cannot be propagated further:  $\sin\phi \in [-1,1]$
- Once the limit is reached, the state-vector and Covariance associated with the last reconstructed track point are “mirrored”:

$$s_{k+1}^- = R s_k^+ \quad P_{k+1}^- = R P_k^+ R^T$$

$$\text{with } R = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & -1 \end{pmatrix}$$

- Finally, the local  $x$  coordinate is propagated by calculating the arch between the two mirrored points:

$$x_{k+1}^- = x_k^+ + \text{arch} * \tan\lambda$$



# ENERGY LOSS CORRECTION

Bethe-Bloch (PDG)  
<https://pdg.lbl.gov/2005/reviews/passagerpp.pdf>

$$\frac{1}{\rho} \frac{dE}{dx} = K \times \frac{Z}{A} \times \frac{z^2}{\beta^2} \left[ \frac{1}{2} \ln \left( \frac{2m_e \gamma^2 \beta^2 T_{max}}{I^2} \right) - \beta^2 - \frac{\delta}{2} \right] \quad [\text{GeV}/(\text{g}/\text{cm}^2)]$$

- Energy loss correction applied to helix fit:

1. Get  $dE/dx$  with Bethe-Bloch and evaluate momentum loss over trajectory in small “momentum-loss” steps
2. Calculate multiplicative factor to update  $q/p_T$ :

$$\frac{q}{p_T} \ast = cP4 = \left( 1 + \frac{\Delta E}{p_{mean}^2} (\Delta E + 2 \times E_{in}) \right)$$

2. Add factor to diagonal element of 5x5 Covariance Matrix  $P$  correspondent to  $q/p_T$  (found through error propagation):

$$P[4][4] += \left( \frac{\sigma_E}{p_{mean}^2} \times \frac{q}{p_T} \right)^2$$

- **Note 1:** These formulas are the same as the ones used by Geant4
- **Note 2:** Applied to both Kalman Filter “step-by-step” and Seeding “globally”

# MS CORRECTION

Molière Formula (PDG)  
<https://pdg.lbl.gov/2005/reviews/passagerpp.pdf>

$$\theta_0 = \frac{13.6 \text{ MeV}}{\beta p} z \sqrt{x/X_0} [1 + 0.038 \ln(x/X_0)]$$

- Multiple Scattering correction applied to Helix fit:
  1. Calculate width of the angular gaussian distribution produced by MS:  $\theta_0$  from Molière formula
  2. Propagate the error to the relevant Helix parameters, obtaining their respective  $\sigma$ 's ( $\sigma_{\sin\phi}$ ,  $\sigma_{\tan\lambda}$ ,  $\sigma_{q/p_T}$ )
  3. Update covariance matrix diagonal elements:

$$\begin{cases} P[2][2] += \sigma_{\sin\phi}^2 \\ P[3][3] += \sigma_{\tan\lambda}^2 \\ P[4][4] += \sigma_{q/p_T}^2 \end{cases}$$

- **Note 1:** These formulas are the same as the ones used by Geant4
- **Note 2:** Applied to both Kalman Filter “step-by-step” and Seeding “globally”

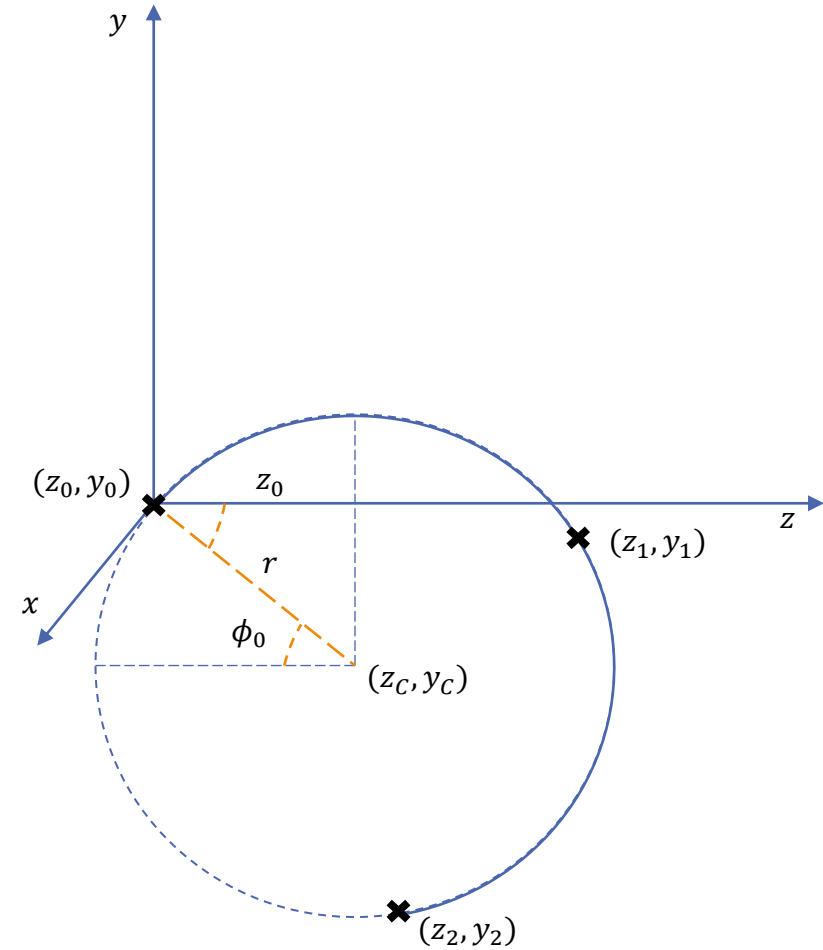


# GLOBAL HELIX FIT AND INITIAL COVARIANCE ESTIMATION

- Seeding for Kalman done with simple 3-point helix fit:
  - $c = 1/r$  and  $\sin \phi_0$  estimated by finding  $(z_c, y_c)$  and  $r$  of the  $yz$  plane circumference:

$$c = 1/r$$

$$\sin \phi_0 = \frac{z_0}{r}$$



# GLOBAL HELIX FIT AND INITIAL COVARIANCE ESTIMATION

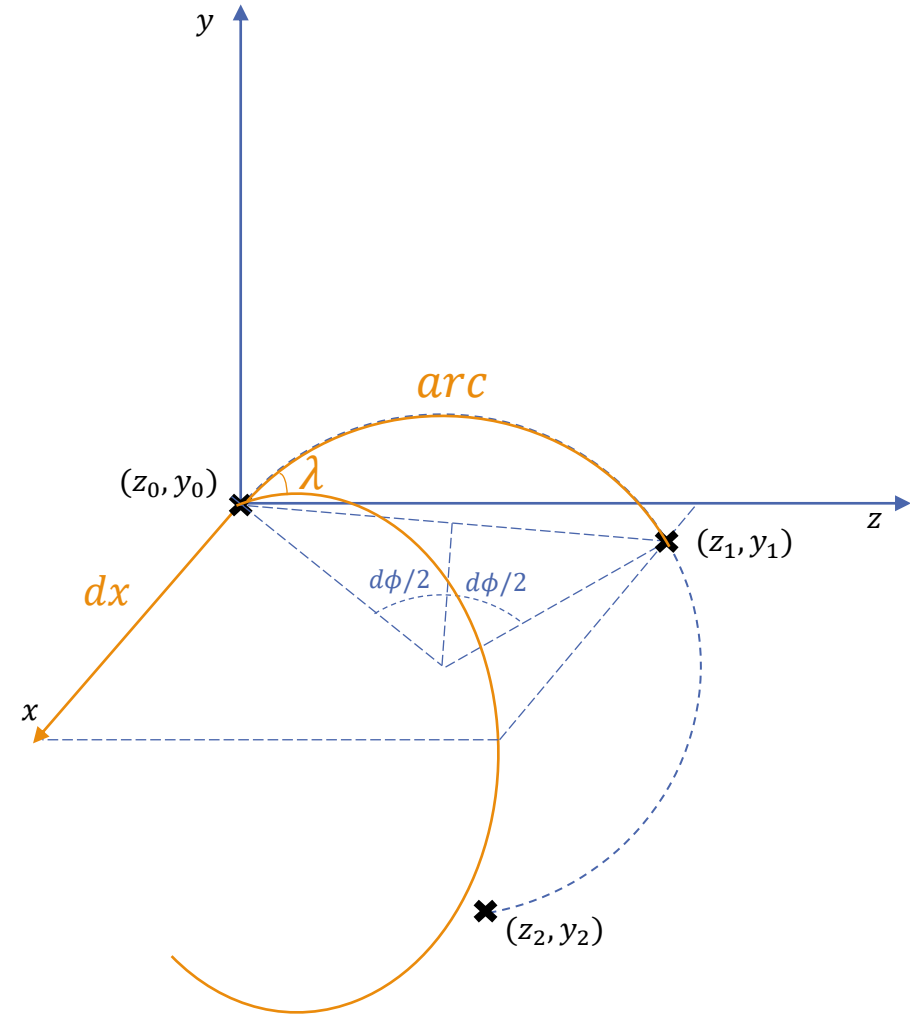
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  - $c = 1/r$  and  $\sin \phi_0$  estimated by finding  $(z_c, y_c)$  and  $r$  of the  $yz$  plane circumference:

$$c = 1/r \quad \sin \phi_0 = \frac{z_0}{r}$$

- $\tan \lambda$  from the  $yz$  plane arc between the first two points and the correspondent movement in the  $x$  direction:

$$\tan \lambda = \frac{dx}{arc} = \frac{dx}{d\phi * r}$$

- Note:** Energy loss and MS corrections applied similarly to Kalman Filter



# ENERGY LOSS AND MS

# ENERGY LOSS: BETHE-BLOCH FORMULA

Bethe-Bloch (PDG)

<https://pdg.lbl.gov/2005/reviews/passagerpp.pdf>

$$\frac{1}{\rho} \frac{dE}{dx} = K \times \frac{Z}{A} \times \frac{z^2}{\beta^2} \left[ \frac{1}{2} \ln \left( \frac{2m_e c^2 \gamma^2 \beta^2 W_{max}}{I^2} \right) - \beta^2 - \frac{\delta}{2} \right] \quad [\text{GeV}/(\text{g}/\text{cm}^2)]$$

- $\rho = 1.032 \text{ g}/\text{cm}^3$
- $K = 4\pi N_A r_e^2 m_e c^2 = 0.307 \text{ 075 MeV mol}^{-1} \text{cm}^2$
- $Z/A = 0.54141 \text{ mol}/\text{g}$
- $z$
- $m_e c^2 = 0.511 \text{ MeV}$
- $W_{max} = 2m_e c^2 \beta^2 \gamma^2$
- $I = 64.7 \times 10^{-9} \text{ GeV}$

Plastic scintillator density

Bethe Bloch constant coefficient

Mean atomic number/mass of plastic scintillator

Atomic number of incident particle

Mass of electron

Low energy approximation of maximum energy transfer

Mean excitation energy

$$\frac{\delta}{2} = \begin{cases} 0 & \ln \beta\gamma < 2.303x_0 \\ \ln \beta\gamma - 1/2 C & \ln \beta\gamma > 2.303x_1 \\ \ln \beta\gamma - 1/2 C + (1/2 C - 2.303X_0) \times \left( \frac{2.303X_1 - \ln \beta\gamma}{2.303(X_1 - X_0)} \right)^3 & \ln \beta\gamma \in [2.303x_0, 2.303x_1] \end{cases}$$

DENSITY  
CORRECTION

with  $C = 2 - \ln \left( \frac{28.816 \times 10^{-9} \sqrt{\rho(Z/A)}}{I} \right)$

$x_0 = 0.1469 \quad x_1 = 2.49$

1st and 2nd junction points for plastic scintillator

# ENERGY LOSS CORRECTION

Bethe-Bloch (PDG)

<https://pdg.lbl.gov/2005/reviews/passagerpp.pdf>

$$\frac{1}{\rho} \frac{dE}{dx} = K \times \frac{Z}{A} \times \frac{z^2}{\beta^2} \left[ \frac{1}{2} \ln \left( \frac{2m_e c^2 \gamma^2 \beta^2 W_{max}}{I^2} \right) - \beta^2 - \frac{\delta}{2} \right] \quad [\text{GeV}/(\text{g}/\text{cm}^2)]$$

- Step by step procedure:

1. Convert into:  $dp/dx = dE/dx \times \beta^{-1}$
2. Calculate number of steps:  $n_{steps} = 1 + (dp/dx \times \Delta x)/step$  with  $step = 0.005$
3. Calculate step-wise total momentum loss:  $\Delta p_{tot} = \sum_{i=0}^{n_{steps}} \Delta p_i = \sum_{i=0}^{n_{steps}} \frac{dp}{dx_i} \Delta x_i$
4. Calculate total energy loss  $\Delta E = E_{in} - \sqrt{p_{out}^2 + m^2}$  with  $p_{out} = p_{in} - \Delta p_{tot}$
5. Apply multiplicative factor:

$$\frac{q}{p_T} *= cP4 = \left( 1 + \frac{\Delta E}{p_{mean}^2} (\Delta E + 2 \times E_{in}) \right)$$

6. Apply correction to covariance matrix:

$$P[4][4] += \left( \frac{\sigma_E}{p_{mean}^2} \times \frac{q}{p_T} \right)^2$$

# KALMAN FILTER: MS CORRECTION

Molière Formula (PDG)

<https://pdg.lbl.gov/2005/reviews/passagerpp.pdf>

$$\theta_0 = \frac{13.6 \text{ MeV}}{\beta p} z \sqrt{x/X_0} [1 + 0.038 \ln(x/X_0)]$$

- $X_0 = 42.54 \text{ cm}$  Radiation length of plastic scintillator in cm
- $x$  is the step length
- $z$  is the charge of incident particle
- Formulas for propagated  $\sigma$ 's:

$$\left\{ \begin{array}{l} \sigma_{\sin \phi} = \theta_0 \cos \phi \sqrt{1 + \tan^2 \lambda} \\ \sigma_{\tan \lambda} = \theta_0 (1 + \tan^2 \lambda) \\ \sigma_{q/p_T} = \theta_0 \tan \lambda \frac{q}{p_T} \end{array} \right.$$

# KALMAN FILTER: ENERGY LOSS CORRECTION

Bethe-Bloch (PDG)  
<https://pdg.lbl.gov/2005/reviews/passagerpp.pdf>

$$\frac{1}{\rho} \frac{dE}{dx} = K \times \frac{Z}{A} \times \frac{z^2}{\beta^2} \left[ \frac{1}{2} \ln \left( \frac{2m_e \gamma^2 \beta^2 T_{max}}{I^2} \right) - \beta^2 - \frac{\delta}{2} \right] \quad [\text{GeV}/(\text{g}/\text{cm}^2)]$$

- Energy loss correction:
  1. Use multiplicative factor  $cP4$  (see slide 7) to update  $q/p_T$
  2. Add factor to diagonal element of 5x5 Covariance Matrix  $P$  correspondent to  $q/p_T$  (found through error propagation):

$$P[4][4] += \left( \frac{\sigma_E}{p_{mean}^2} \times \frac{q}{p_T} \right)^2$$

- **NOTE:**  $\sigma_E = k \times \sqrt{|\Delta E|}$  where  $k$  is a tunable parameter set at 0.07

# KALMAN FILTER: MS CORRECTION

Molière Formula (PDG)  
<https://pdg.lbl.gov/2005/reviews/passagerpp.pdf>

$$\theta_0 = \frac{13.6 \text{ MeV}}{\beta p} z \sqrt{x/X_0} [1 + 0.038 \ln(x/X_0)]$$

- Multiple Scattering smearing simulated in three steps:
  1. Obtain parameter  $\sigma$ 's ( $\sigma_{\sin\phi}$ ,  $\sigma_{\tan\lambda}$ ,  $\sigma_{q/p_T}$ ) through error propagation as described in slide 6
  2. Update covariance matrix diagonal elements:

$$\begin{cases} P[2][2] += \sigma_{\sin\phi}^2 \\ P[3][3] += \sigma_{\tan\lambda}^2 \\ P[4][4] += \sigma_{q/p_T}^2 \end{cases}$$



# KALMAN FILTER

# KALMAN FILTER IN GENERAL

1. Make **a priori predictions** for the current step's state and covariance matrix using the **a posteriori best estimate of the previous step** (i.e. updated using measurement)

STATE VECTOR

$$s_k^- = f(s_{k-1}^+, X_{k-1})$$

COVARIANCE MATRIX

$$P_k^- = F_{k-1} P_{k-1}^+ F_{k-1}^T + Q$$

$$F_{k-1} = \left. \frac{\partial f}{\partial s} \right|_{s_{k-1}^+, X_{k-1}}$$

JACOBIAN

$Q$

PROCESS NOISE  
COVARIANCE

**Note:** In the first iteration step we use step 0 estimates for the state vector and the covariance matrix  $(s_0, P_0)$ , which can be made very roughly

# KALMAN FILTER IN GENERAL

2. Calculate the measurement residual and the Kalman Gain

RESIDUAL

$$\tilde{y}_k = m_k^h - H(s_k^-)$$

KALMAN GAIN

$$K_k = P_k^- H^T (R + H P_k^- H^T)^{-1}$$

$R$

MEASUREMENT  
NOISE COVARIANCE

$H$

CONVERSION  
MATRIX

3. Update the estimate

STATE VECTOR

$$s_k^+ = s_k^- + K_k \tilde{y}$$

COVARIANCE MATRIX

$$P_k^+ = (1 - K_k H) P_k^-$$

**Note:** in the case where  $R$  is a null matrix  $s_k^+ = s_k^h$  and  $P_k^+ = 0$

**Note:** the conversion matrix is needed to make the dimensions of vectors and matrixes turn out right. For example if  $s_k^h$  is a 2-D vector and  $s_k^-$  is 5-D, then  $H$  would be a  $2 \times 5$  matrix:

$$H = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \end{pmatrix}$$

# KALMAN FILTER MODEL

- Use parametrization used in ALICE: free parameter  $z$ , state vector  $s = (y, x, \sin\phi, \tan\lambda, \frac{q}{p_T})$  ( $\phi$  azimuthal angle,  $\lambda$  dip-angle,  $p_T$  transverse momentum in  $yz$  plane), evolution function:

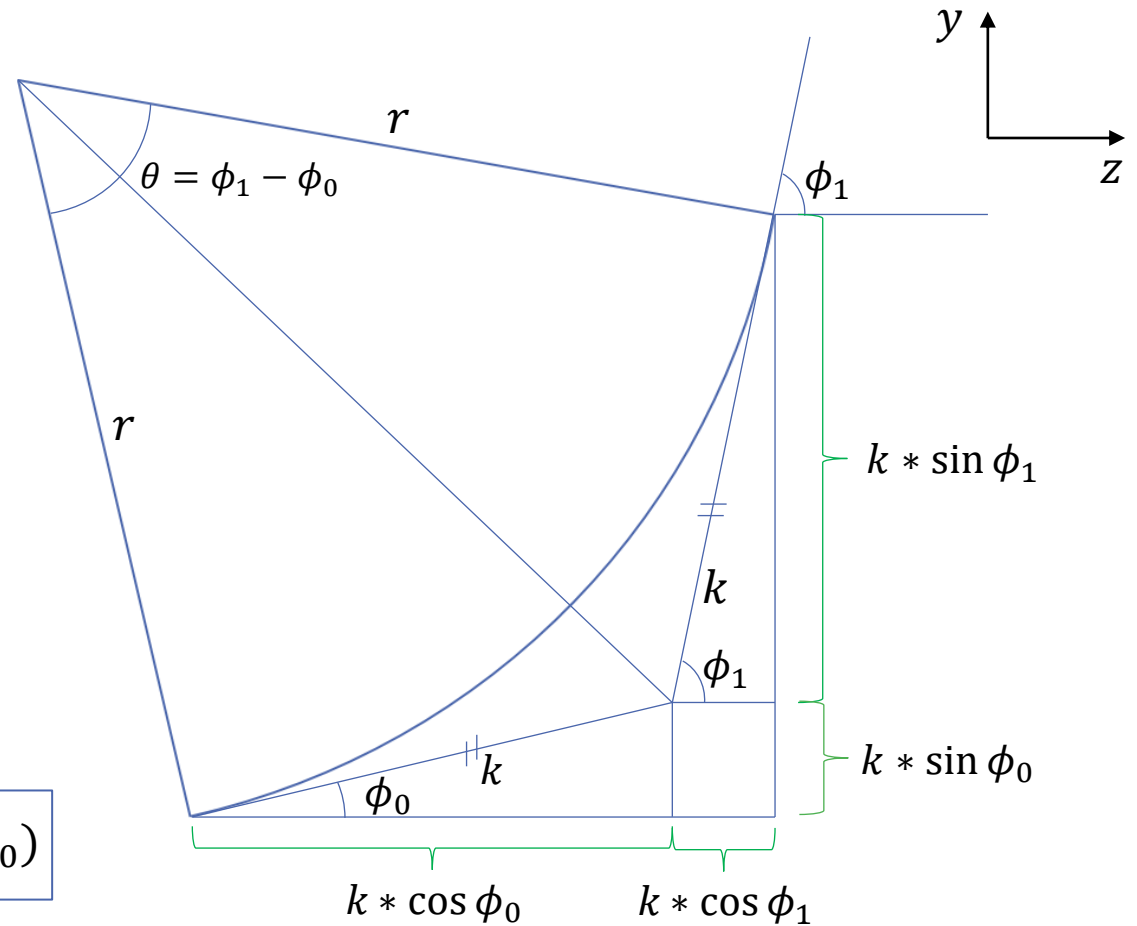
$$0 \quad \frac{dy}{dz} = \frac{k * (\sin\phi_0 + \sin\phi_1)}{k * (\cos\phi_0 + \cos\phi_1)}$$

$$y_1 = y_0 + \frac{(\sin\phi_0 + \sin\phi_1)}{(\cos\phi_0 + \cos\phi_1)} * dz$$

$$1 \quad dx = arch * \tan\lambda = \theta * r * \tan\lambda$$

$$\theta = \phi_1 - \phi_0 = \arcsin(\sin(\phi_1 - \phi_0)) = \arcsin(\cos\phi_0 \sin\phi_1 - \cos\phi_1 \sin\phi_0)$$

$$x_1 = x_0 + \tan\lambda * \frac{r}{q} * \arcsin(\cos\phi_0 \sin\phi_1 - \cos\phi_1 \sin\phi_0)$$



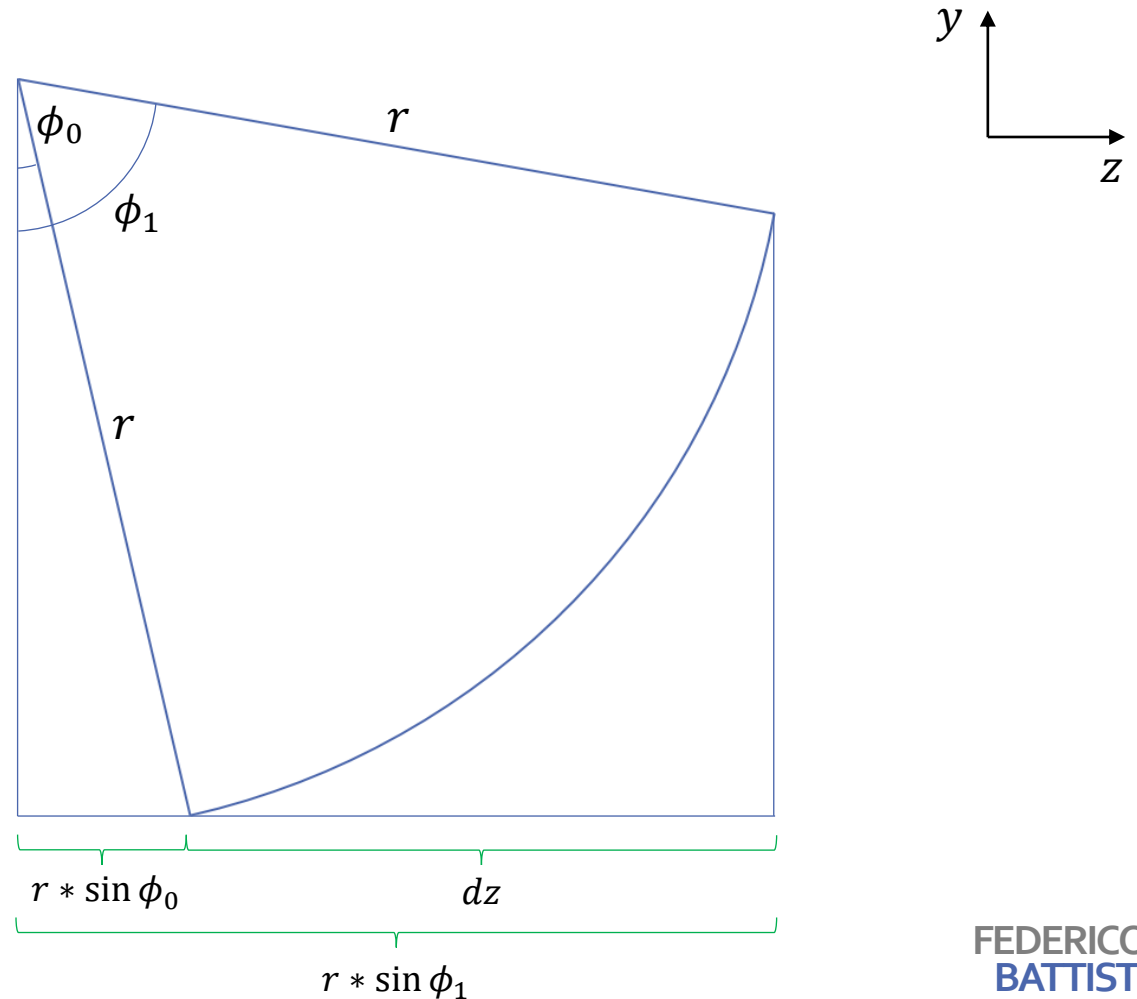
# KALMAN FILTER MODEL

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②  $dz = r * \sin \phi_1 - r * \sin \phi_0$

$$\sin \phi_1 = \sin \phi_0 + \frac{dz}{r}$$

③ & ④ are static



# HELIX FIT

# K-F UPDATE: GLOBAL HELIX FIT AND INITIAL COVARIANCE ESTIMATION

- $c = 1/r$  and  $\sin \phi_0$  estimated by finding  $(z_c, y_c)$  and  $r$  of the  $yz$  plane circumference passing through the first, last and middle hit point of the particle trajectory
- After traslating the coordinate system to have the origin on the first point  $(z_0, y_0) \rightarrow (0,0)$  we have the circumference equations:

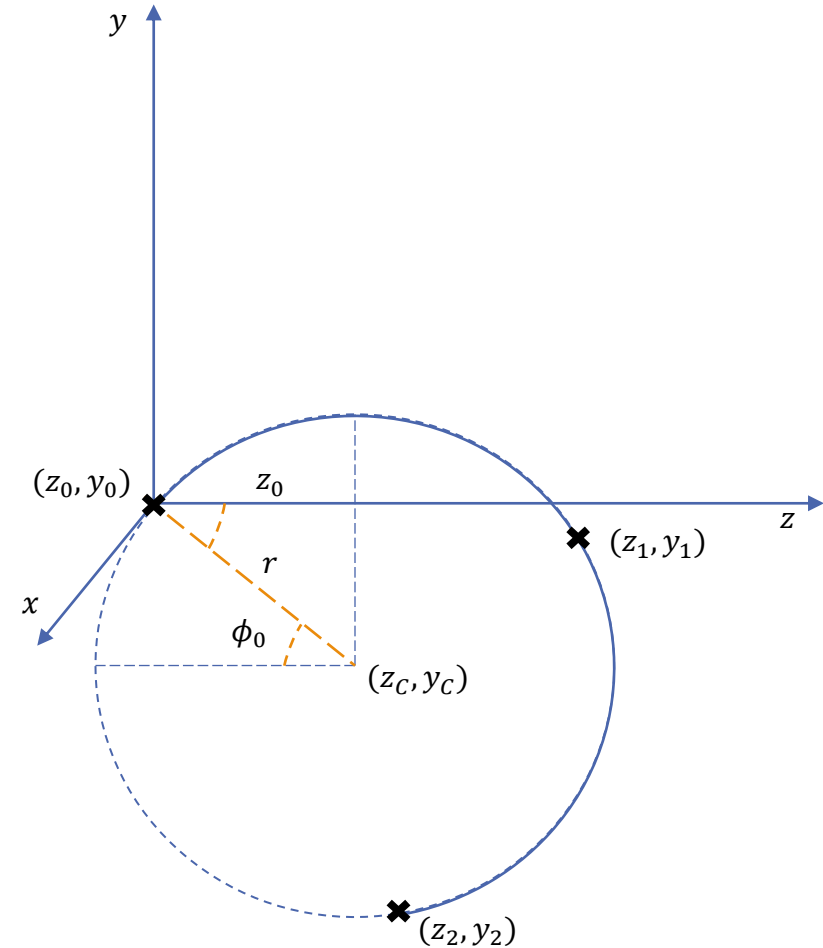
$$\begin{cases} z_c^2 + y_c^2 = r^2 \\ (z_1 - z_c)^2 + (y_1 - y_c)^2 = r^2 \\ (z_2 - z_c)^2 + (y_2 - y_c)^2 = r^2 \end{cases}$$



$$\begin{cases} z_c = \frac{1}{2} \left( z_2 - y_2 \frac{z_1(z_1 - z_2) + y_1(y_1 - y_2)}{z_2 y_1 - z_1 y_2} \right) \\ y_c = \frac{1}{2} \left( z_2 - y_2 \frac{z_1(z_1 - z_2) + y_1(y_1 - y_2)}{z_2 y_1 - z_1 y_2} \right) \\ r = \sqrt{z_c^2 + y_c^2} \end{cases}$$



$$\begin{aligned} c &= 1/r \\ \sin \phi_0 &= \frac{z_0}{r} \end{aligned}$$



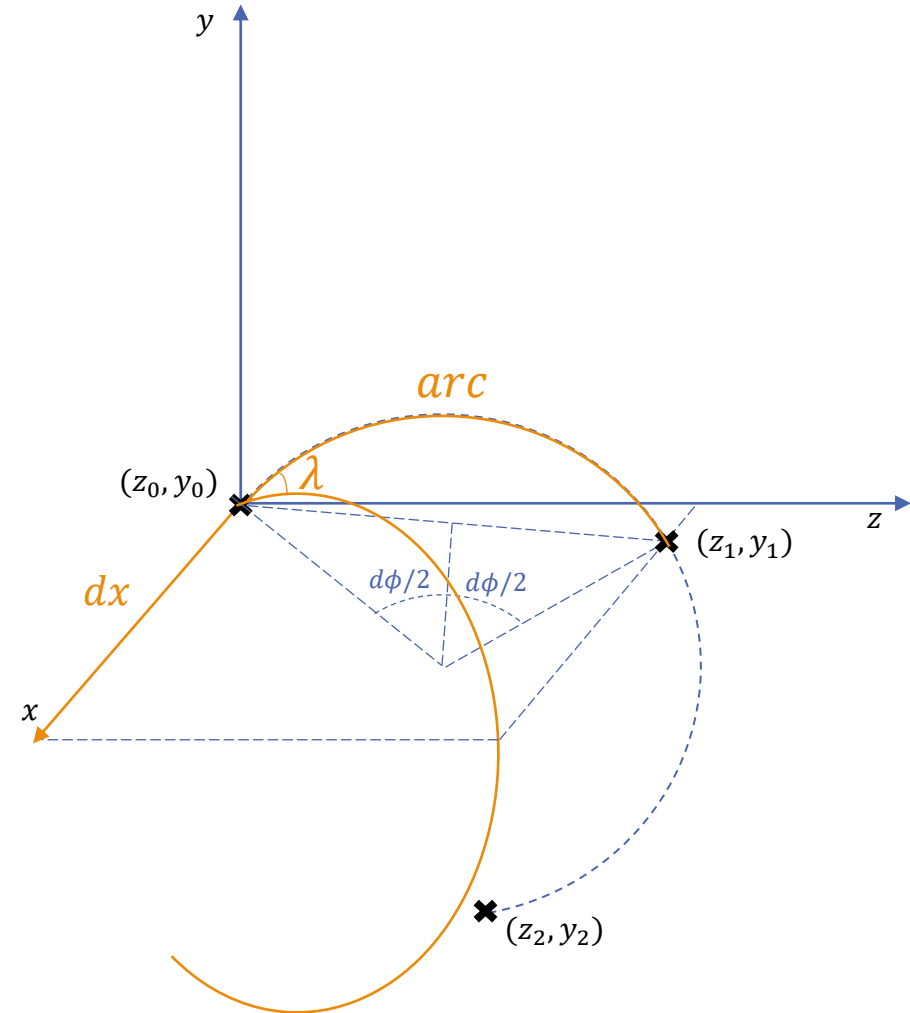
# K-F UPDATE: GLOBAL HELIX FIT AND INITIAL COVARIANCE ESTIMATION

- We evaluate  $\tan \lambda$  from the  $yz$  plane arc between the first two points and the correspondent movement in the  $x$  direction (magnetic field direction) using  $r$  estimate from previous step:

$$d\phi = 2 \arcsin\left(\frac{\text{chord}}{2r}\right)$$
$$= 2 \arcsin\left(\frac{\sqrt{(y_1 - y_0)^2 + (z_1 - z_0)^2}}{2r}\right)$$



$$\tan \lambda = \frac{dx}{arc} = \frac{dx}{d\phi * r}$$





# K-F UPDATE: GLOBAL HELIX FIT AND INITIAL COVARIANCE ESTIMATION

- Given parameter estimation from global helix fit, **estimate uncertainties through error propagation**
- Uncertainties associated with  $x$  and  $y$ :  $\sigma_{xy}$  ;  $z$  free parameter with no uncertainty  $\sigma_z = 0$  (as in the Kalman filter)
- Formula for  **$\sin \phi_0$  estimation** is function of  $f(z_0, y_0, z_1, y_1, z_2, y_2)$  but since  $\sigma_z = 0$ , consider only  $f(y_0, y_1, y_2) \rightarrow$   
**From error propagation we get:**

$$\sigma_{\sin \phi_0} = \sqrt{\left(\frac{\partial f(y_0, y_1, y_2)}{\partial y_0}\right)^2 \sigma_{xy}^2 + \left(\frac{\partial f(y_0, y_1, y_2)}{\partial y_1}\right)^2 \sigma_{xy}^2 + \left(\frac{\partial f(y_0, y_1, y_2)}{\partial y_2}\right)^2 \sigma_{xy}^2}$$

- This can be approximated as:

$$\sigma_{\sin \phi_0} = \sqrt{\left(\frac{f(y_0 + \sigma_{xy}, y_1, y_2)}{\sigma_{xy}}\right)^2 \sigma_{xy}^2 + \left(\frac{f(y_0, y_1 + \sigma_{xy}, y_2)}{\sigma_{xy}}\right)^2 \sigma_{xy}^2 + \left(\frac{f(y_0, y_1, y_2 + \sigma_{xy})}{\sigma_{xy}}\right)^2 \sigma_{xy}^2}$$

# K-F UPDATE: GLOBAL HELIX FIT AND INITIAL COVARIANCE ESTIMATION

- Repeat the process with other parameters to get respective uncertainties
- Estimate for covariance matrix  $P_0$  is diagonal matrix with:

$$P_0 = \begin{pmatrix} \sigma_{xy}^2 & 0 & 0 & 0 & 0 \\ 0 & \sigma_{xy}^2 & 0 & 0 & 0 \\ 0 & 0 & \sigma_{\sin\phi}^2 & 0 & 0 \\ 0 & 0 & 0 & \sigma_{\tan\lambda}^2 & 0 \\ 0 & 0 & 0 & 0 & \sigma_{q/p_T}^2 \end{pmatrix}$$

- **Note:** off-diagonal elements could also be calculated, but are not at the moment