

# $A_N$ vs. $P_T$ Unfolding

**RadLab Group Meeting**

**2020/07/22**

**9:00 PM (KST/JST)**

**Benard Mulilo (KU/RIKEN)**

# Asymmetry and Parameter Scanning – Functional Forms

For different MC samples (OPE, Pythia6, Pythia8, Dpmjet and UPC)

- Spin and  $P_T$  dependent weights are scanned over a wide range of linear, quadratic and cubic parts of true  $A_N$ s using random generator.

Three different functional forms have been used for the weights:

- ❖ 3rd order polynomial ( $w_{pol3}$ ) based weight:

$$w_{pol3} = 1 + (a + b * P_{T,T} + c * P_{T,T}^2 + d * P_{T,T}^3) * \cos(\Phi_{T,T} + spin * \pi)$$

- ❖ Power function ( $w_{pow}$ ) based weight:

$$w_{pow} = 1 + (a + b * P_{T,T}^c) * \cos(\Phi_{T,T} + spin * \pi)$$

- ❖ Exponential function ( $w_{exp}$ ) based weight:

$$w_{exp} = 1 + \left( a + b * \left( 1 - \exp \left( (c + d) * P_{T,T} \right) \right) \right) * \cos(\Phi_{T,T} + spin * \pi)$$

a = constant term (not useful – can not distinguish left-right asymmetry)

b = linear part scanned over a wide range  $b_{min} < b < b_{max}$

c = quadratic part scanned over a wide range between  $c_{min} < c < c_{max}$

d = cubic part scanned over a wide range between  $d_{min} < d < d_{max}$

s = spin ( $\uparrow\downarrow$ )

$P_{T,T}$  = true transverse momentum and  $\Phi_{T,T}$  = true azimuthal angle distributions

# Asymmetry and Parameter Scanning – Chi-Square (Min)

Calculate Chi-Square based on the reconstructed asymmetries and the experimental data asymmetries (Minjung) for the three different functional forms:

1. Third order polynomial (pol3):

$$\chi_{pol3}^2 = \sum_i \frac{(A_{N,i}^{Minjung} - A_{N,i}^{w_{pol3},reco})^2}{\Delta A_{N,i}^{2,Minjung} + \Delta A_{N,i}^{2,w_{pol3},reco}}$$

2. Power function (pow)

$$\chi_{pow}^2 = \sum_i \frac{(A_{N,i}^{Minjung} - A_{N,i}^{w_{pow},reco})^2}{\Delta A_{N,i}^{2,Minjung} + \Delta A_{N,i}^{2,w_{pow},reco}}$$

3. Exponential function.

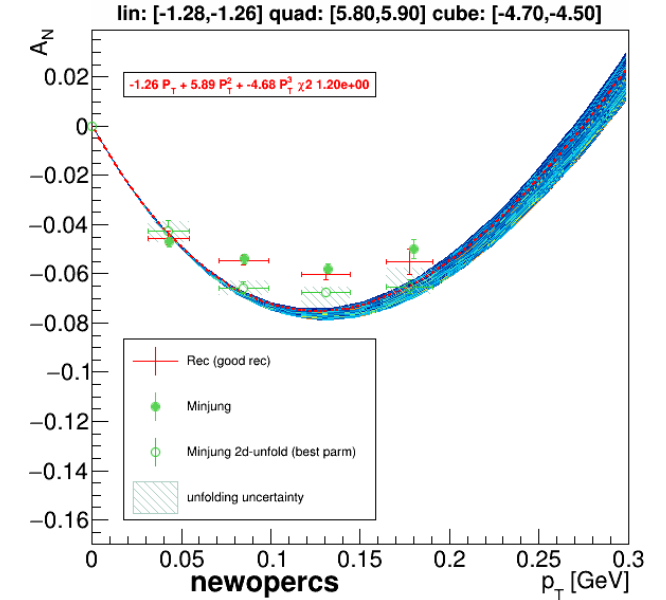
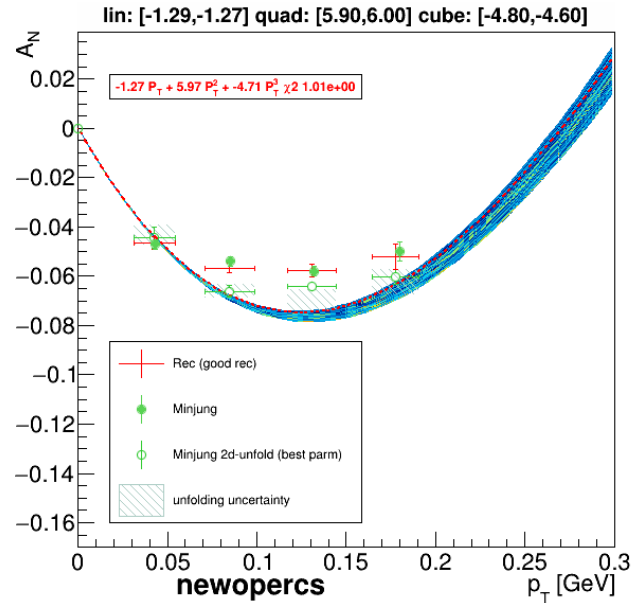
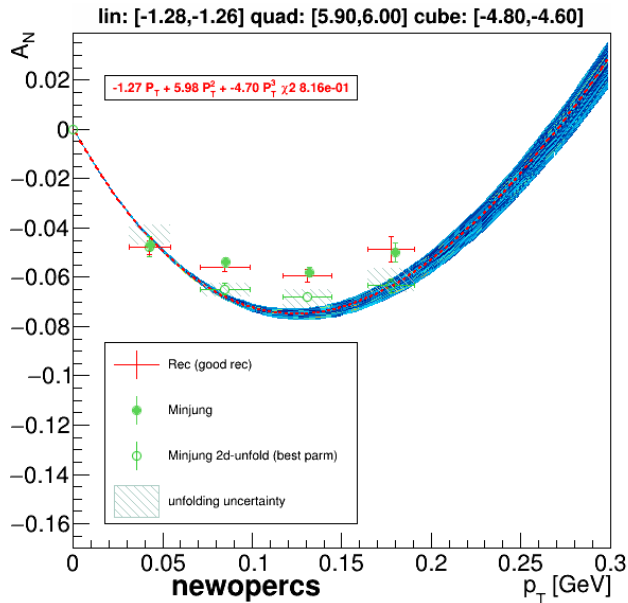
$$\chi_{exp}^2 = \sum_i \frac{(A_{N,i}^{Minjung} - A_{N,i}^{w_{exp},reco})^2}{\Delta A_{N,i}^{2,Minjung} + \Delta A_{N,i}^{2,w_{exp},reco}}$$

Best parameter is found (i.e. parameter with lowest Chi-Square for each functional form).

# Asymmetry based on 3rd order polynomial function

- Pol3 weights depending on spin and pt are scanned over a wide range of linear, quadratic and cubic parts of true  $A_N$ 's.

$$W_{pol3} = 1 + (a + b * P_{T,T} + c * P_{T,T}^2 + d * P_{T,T}^3) * \cos(\Phi_{T,T} + spin * \pi)$$



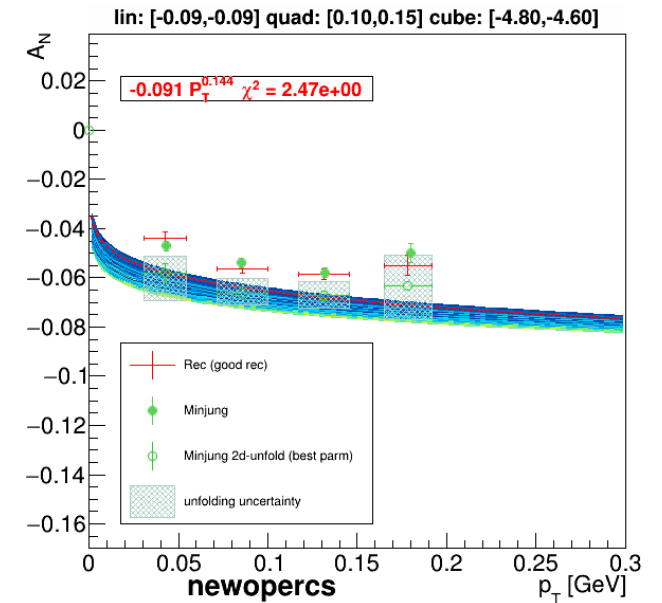
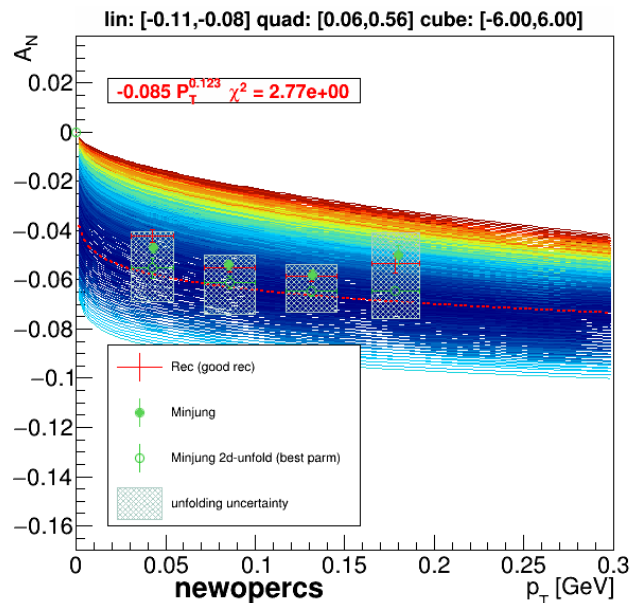
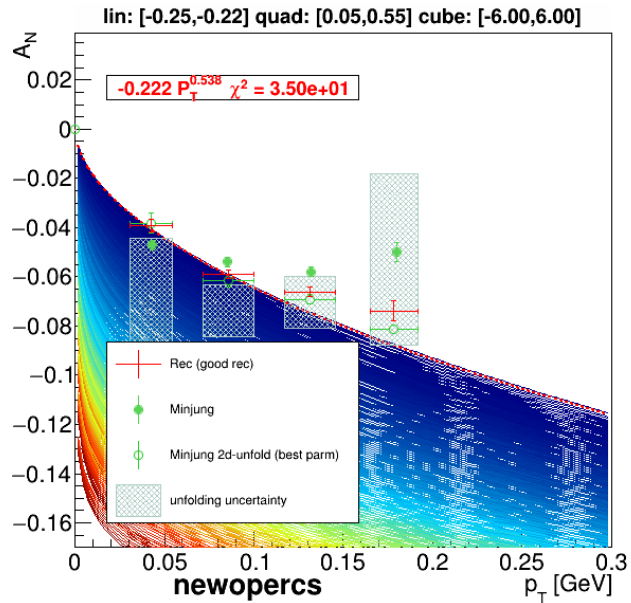
- Based on 3rd degree polynomial weighted reconstructed asymmetries and Minjung's result, Chi-Square is calculated.

$$\chi_{pol3}^2 = \sum_i \frac{(A_{N,i}^{Minjung} - A_{N,i}^{W_{pol3},reco})^2}{\Delta A_{N,i}^{2,Minjung} + \Delta A_{N,i}^{2,W_{pol3},reco}}$$

- Best parameter is found (i.e. lowest  $\chi^2$ ).
- Unfold using best parametrization corresponding to lowest  $\chi^2$

$\chi^2$  based on 3rd order polynomial weighted  $A_N$  (reco) and Minjung's  $A_N$ 's

# Asymmetry weighted based on the power function



- Power function weights depending on spin and  $p_T$  are scanned over a wide range of linear, quadratic and cubic parts of true  $A_N$ 's.

$$w_{pow} = 1 + (a + b * P_{T,T}^c) * \cos(\Phi_{T,T} + spin * \pi)$$

Power weight

- Based on exponentially weighted reconstructed asymmetries and Minjung's result, Chi-Square is calculated.

$$\chi_{pow}^2 = \sum_i \frac{(A_{N,i}^{Minjung} - A_{N,i}^{w_{pow},reco})^2}{\Delta A_{N,i}^{2,Minjung} + \Delta A_{N,i}^{2,w_{pow},reco}}$$

- Best parameter is found (i.e. lowest  $\chi^2$ ).
- Unfold using best parametrization corresponding to lowest  $\chi^2$ .

$\chi^2$  based on power weighted  $A_N$  (reco) and Minjung's  $A_N$ 's

# Asymmetry weighted based on the exponential function

- Exponential weights depending on spin and pt are scanned over a wide range of linear, quadratic and cubic parts of true  $A_N$ 's.

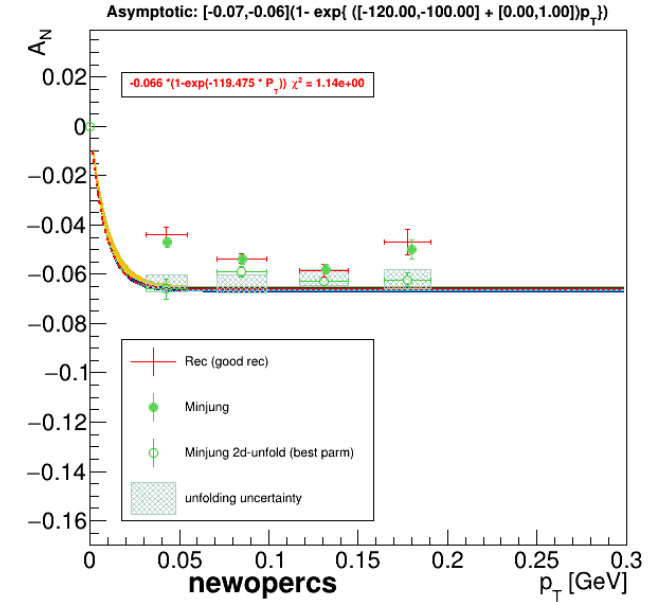
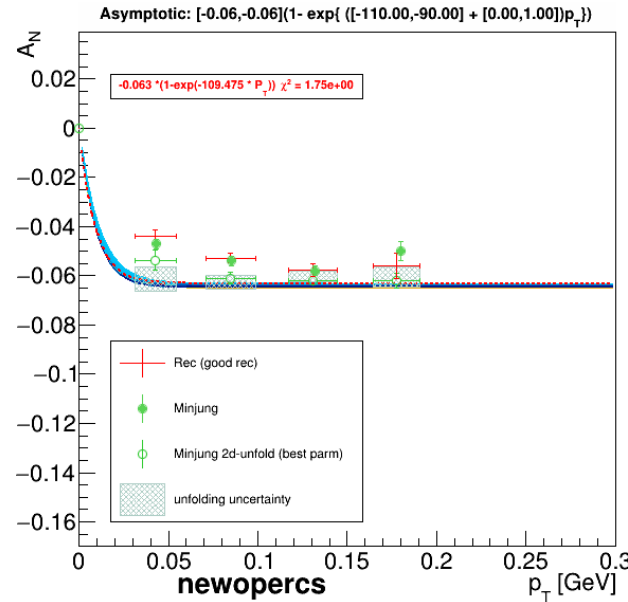
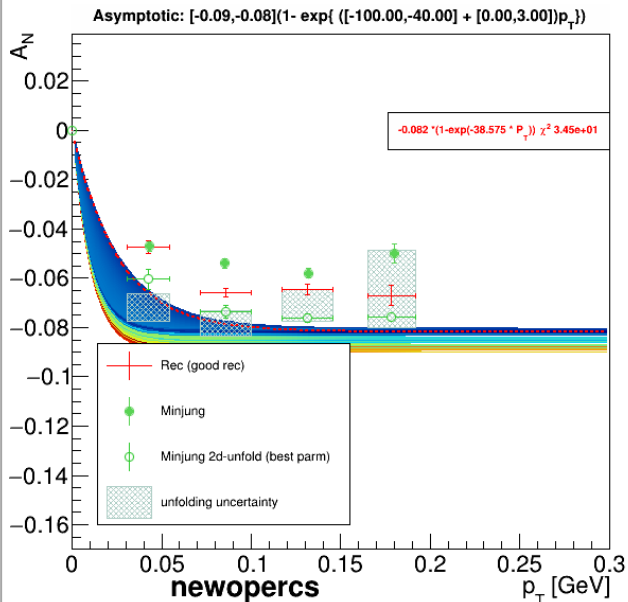
Exponential weight  $W_{exp} = 1 + \left( a + b * \left( 1 - \exp \left( (c + d) * P_{T,T} \right) \right) \right) * \cos(\Phi_{T,T} + spin * \pi)$

- Based on exponentially weighted reconstructed asymmetries and Minjung's result, Chi-Square is calculated.

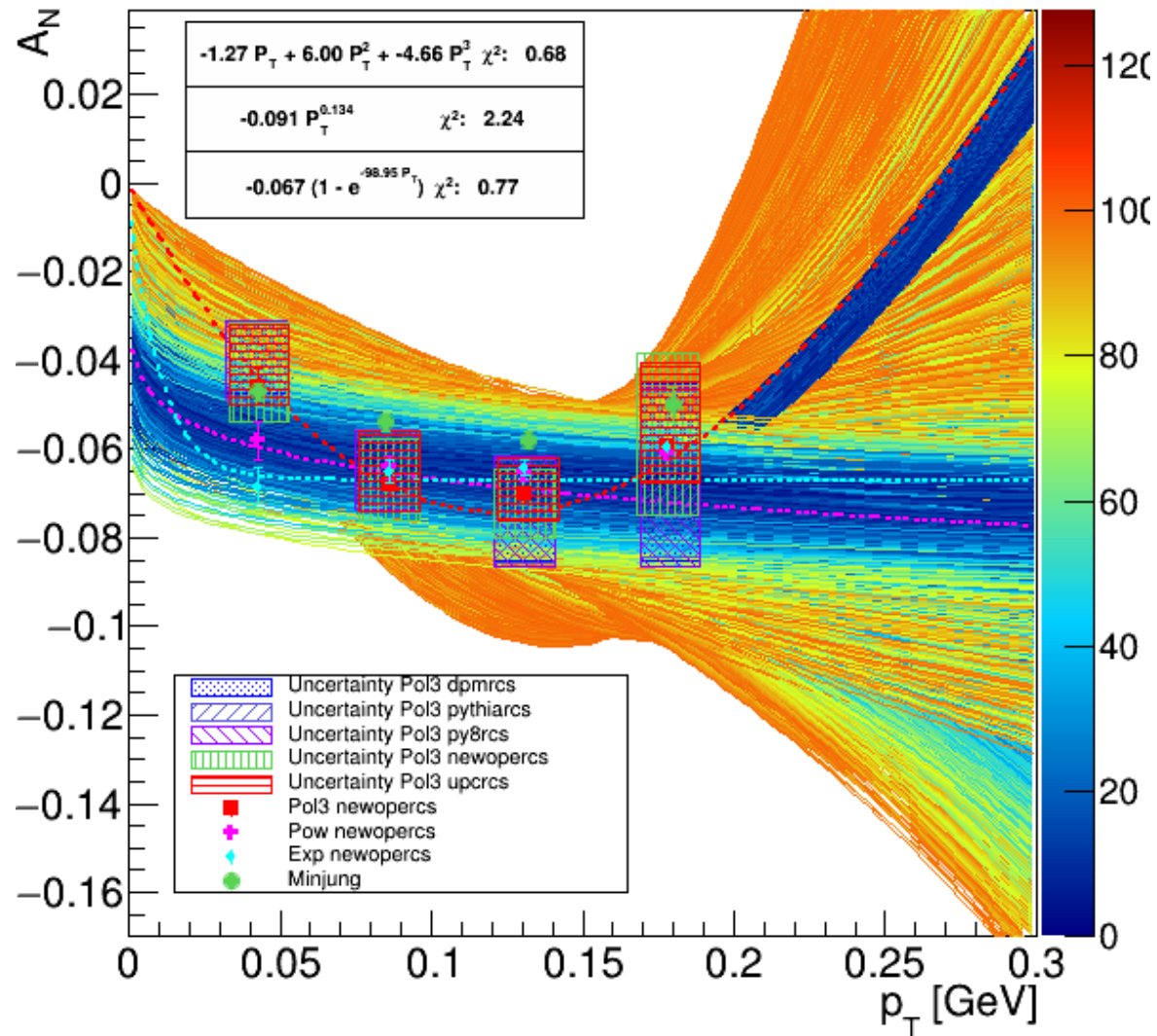
$$\chi^2_{exp} = \sum_i \frac{(A_{N,i}^{Minjung} - A_{N,i}^{W_{exp},reco})^2}{\Delta A_{N,i}^{2,Minjung} + \Delta A_{N,i}^{2,W_{exp},reco}}$$

$\chi^2$  based on exponentially weighted  $A_N$  (reco) and Minjung's  $A_N$ 's

- Best parameter is found (i.e. lowest  $\chi^2$ ).
- Unfold using best parametrization corresponding to lowest  $\chi^2$



# Asymmetry based on Polynomial, Power and Exponential Functions



Files range: 0 and 100 (different parameters & Chi2)

Best chosen parameters:

➤ Pol3 Chi2 = 0.68

➤ Powr Chi2 = 2.24

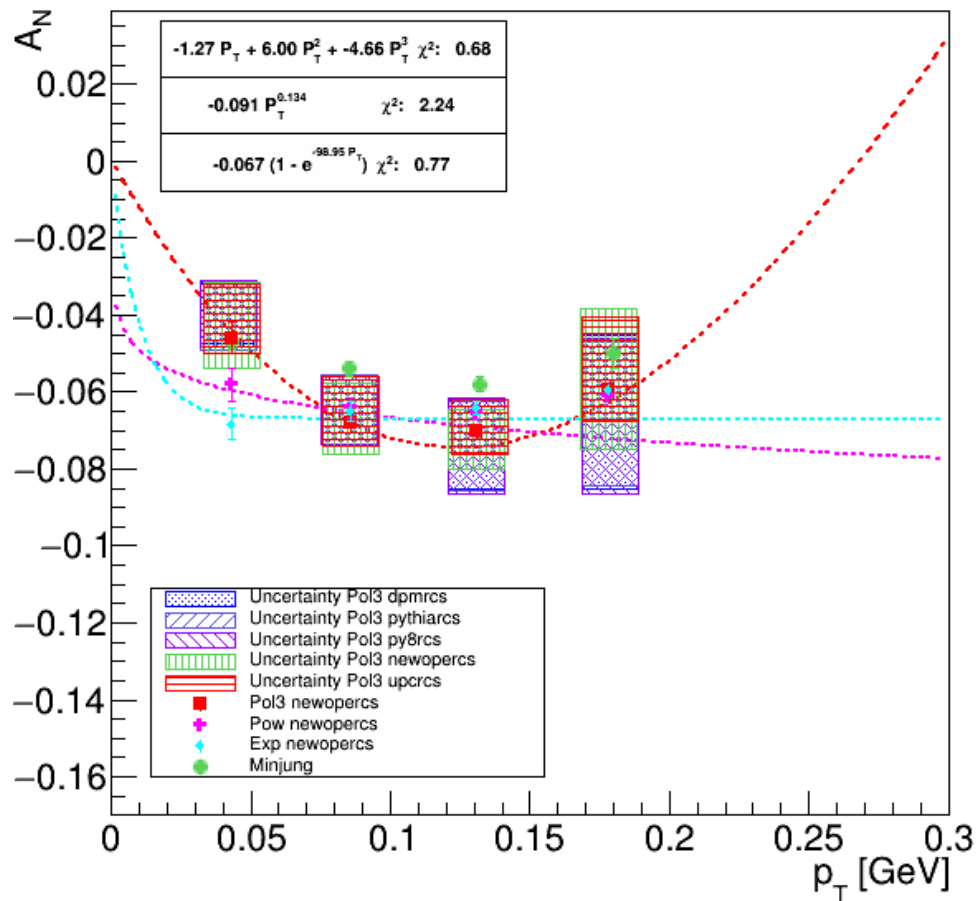
➤ Expo Chi2 = 0.77

All show reasonable  $\chi^2$  with respect to Minjung's asymmetry results.

All MC uncertainties:

➤ Shown for Pol3 Dpmjet, Pythia6, Pythia8, Ope & Upc.

# Inclusive Result with Statistical Uncertainties from Monte Carlo only



Files range: 0 and 100 (different parameters & Chi2)

Best chosen parameters:

➤ Pol3 Chi2 = 0.68

➤ Powr Chi2 = 2.24

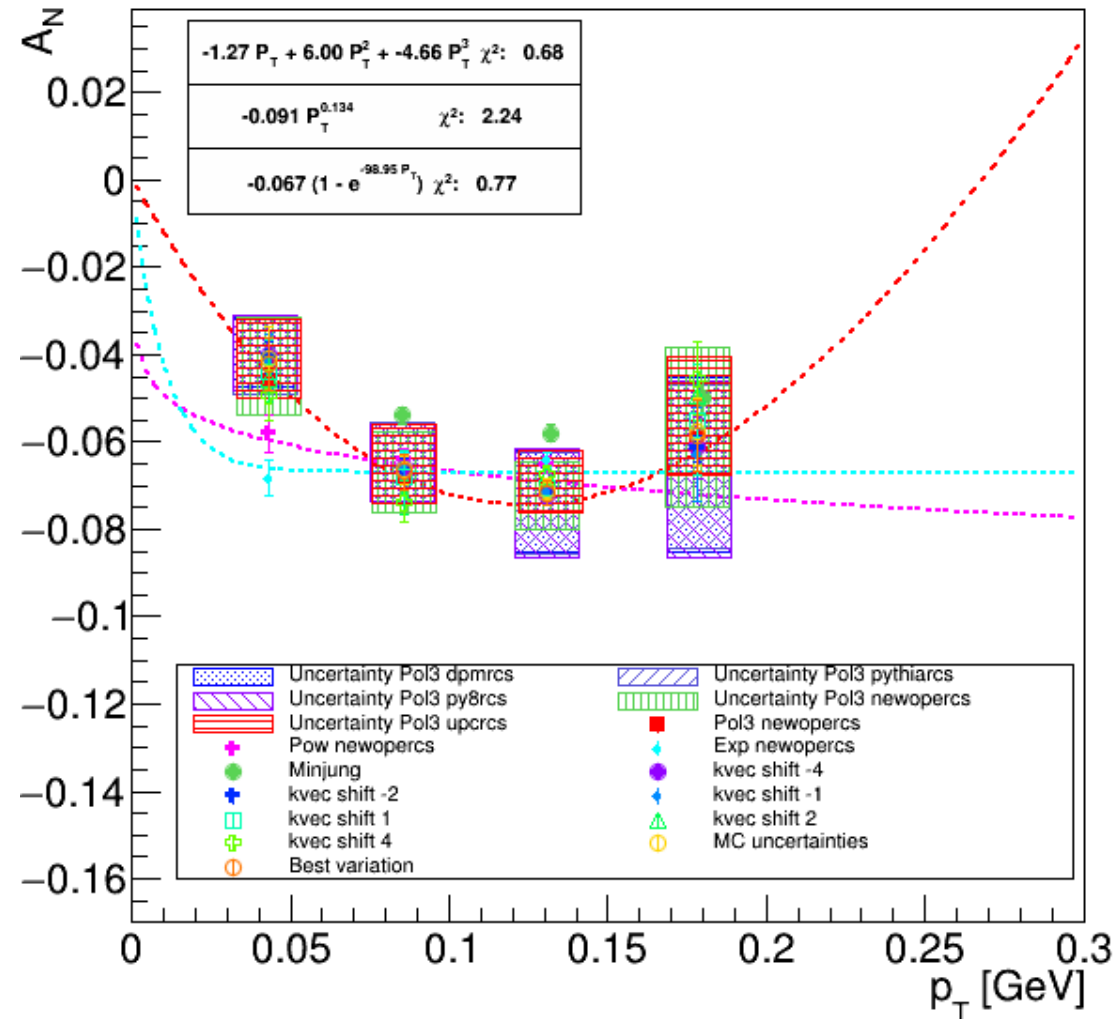
➤ Expo Chi2 = 0.77

MC (Poly3) systematic uncertainties:

➤ For Dpmjet, Pythia6, Pythia8, Ope & Upc using Pol3.



# Inclusive Result with Statistical Uncertainties from MC and Unfolding



Files range: 0 and 100 (different parameters & Chi2)

Best parameters:

➤ Pol3 Chi2 = 0.68

➤ Powr Chi2 = 2.24

➤ Expo Chi2 = 0.77

All systematic uncertainties:

1. MC (Pol3): Dpmjet, Pythia6, Pythia8, Ope & Upc.
2. Unfolding: (Regularization: kvec shifts,  $\pm 1$ ,  $\pm 2$ ,  $\pm 4$ )
3. Uncertainties on response matrix (MC uncertainties)
4. Best variation

# Summary:

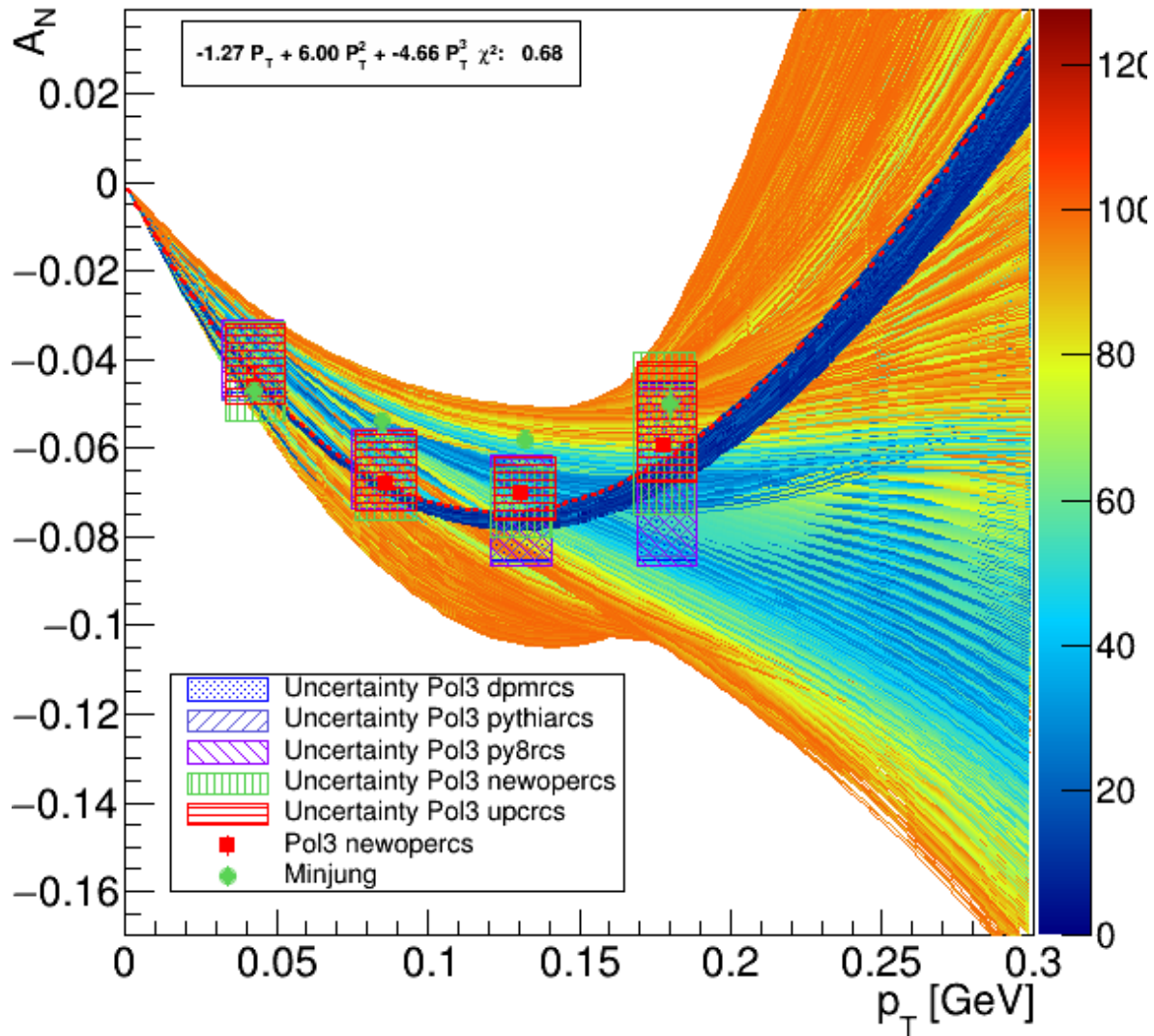
- Unfolding is reasonable with best parametrization.

Uncertainties studied through:

- Varying best regularization parameter of the unfolding.
- MC uncertainties through response matrix
- Comparison of different MC samples
- Repeating of best parametrization

**BACKUP**

# Asymmetry based on Polynomial, Power and Exponential Functions



Files range: 0 and 100 (different parameters & Chi2)

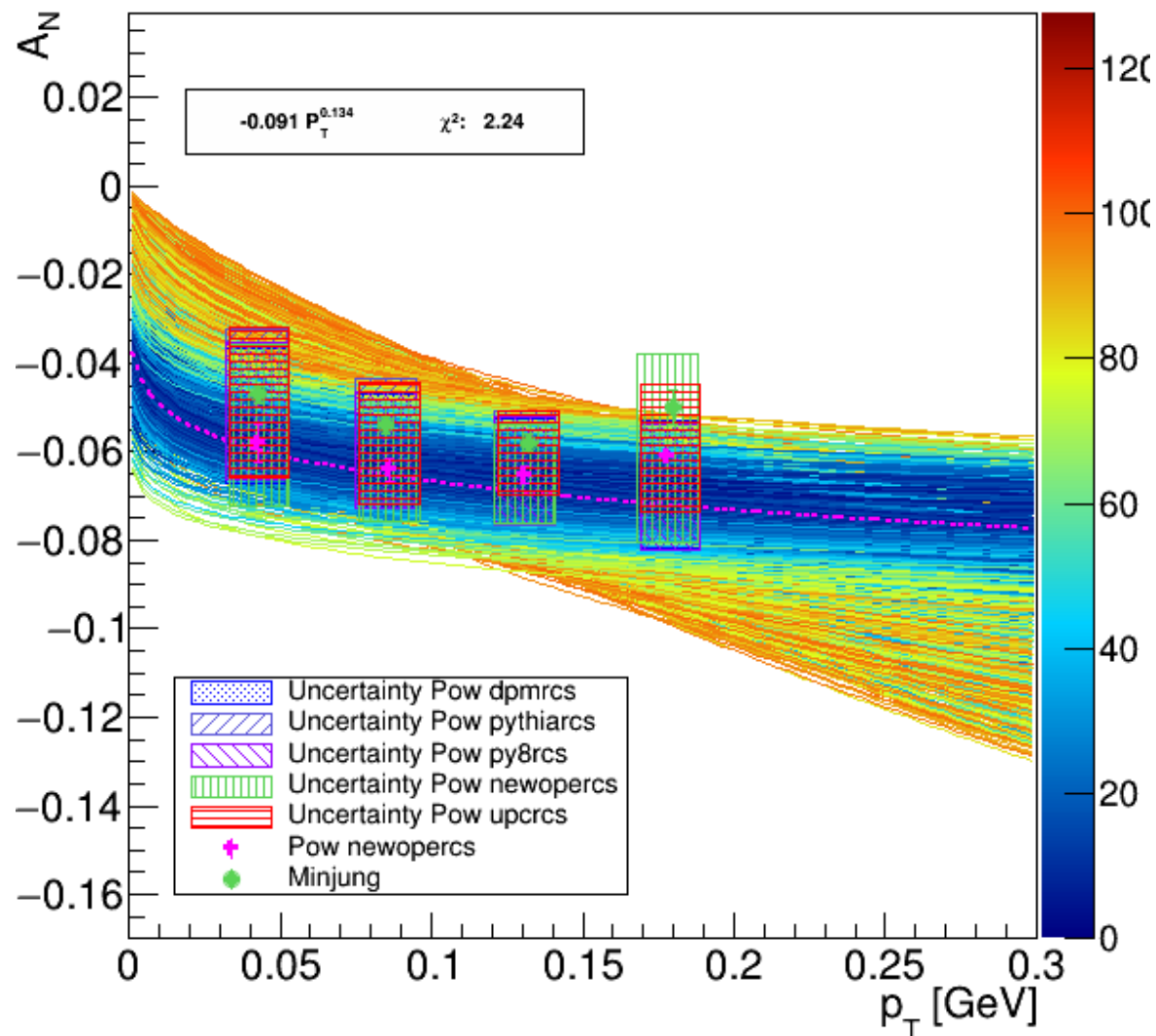
Best parameters:

- Linear = -1.27
- Quadratic = +6.00
- Cubic = -4.66
- Chi-Square = 0.68

Uncertainties:

- Shown for Pol3 Dpmjet, Pythia6, Pythia8, Ope & Upc.

# Asymmetry based on Polynomial, Power and Exponential Functions



Files range: 0 and 100 (different parameters & Chi2)

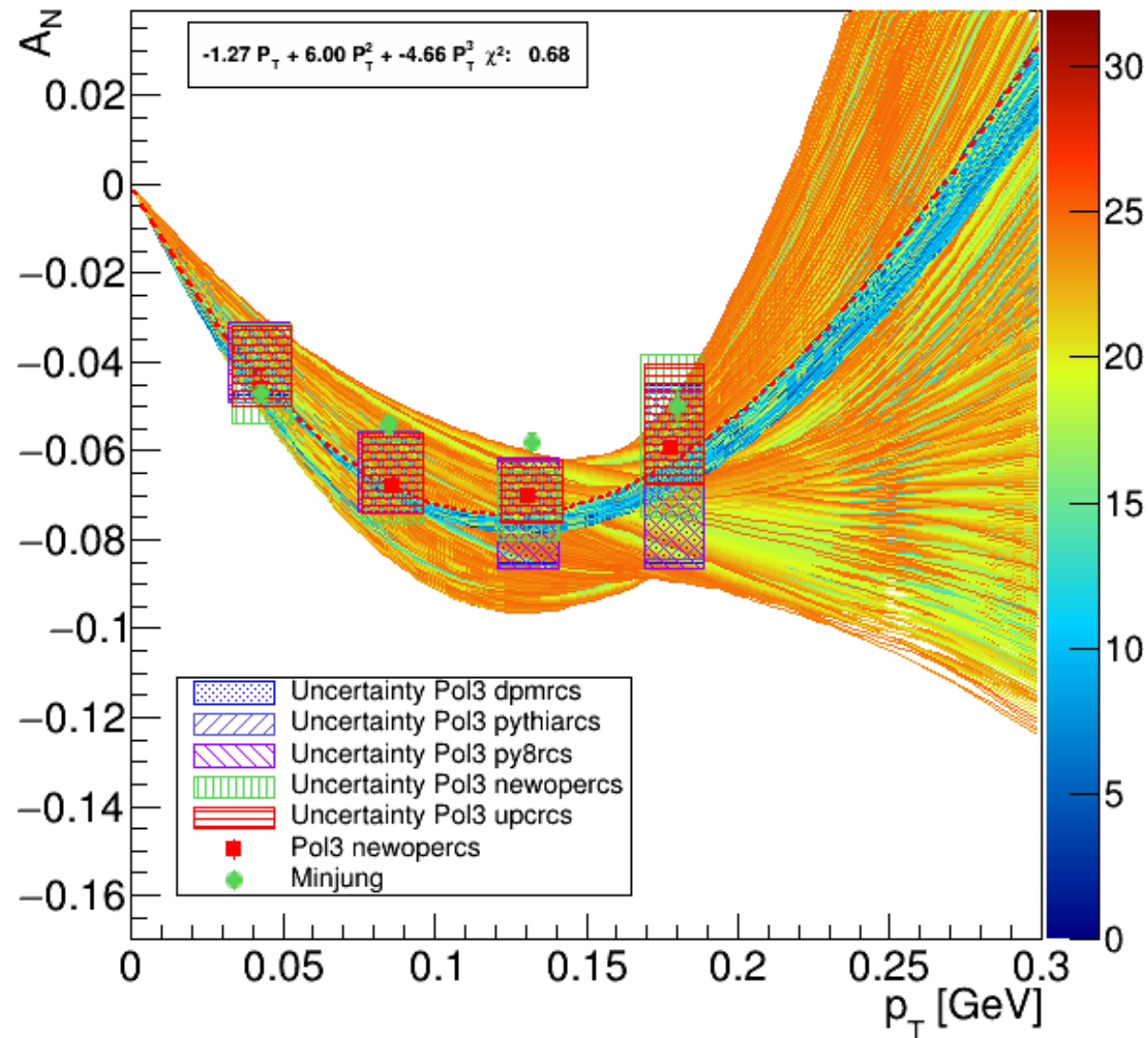
Best chosen parameters:

- Linear = -0.091
- Quadratic = +0.134
- Chi-Square = 2.24

Uncertainties:

- Shown for Pol3 Dpmjet, Pythia6, Pythia8, Ope & Upc.

# Asymmetry based on Polynomial, Power and Exponential Functions



Files range: 0 and 25 (different parameters & Chi2)

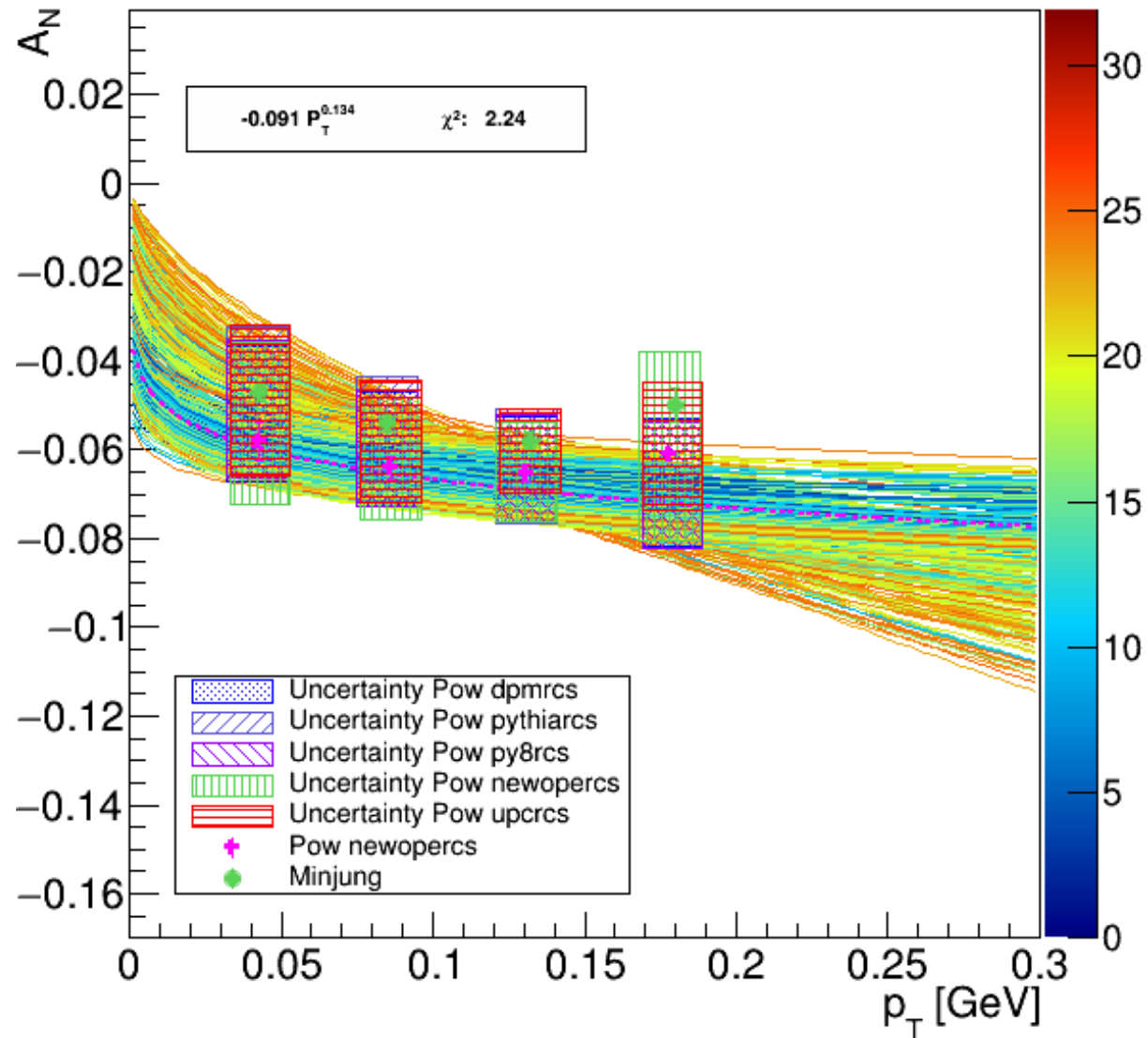
Best parameters:

- Linear = -1.27
- Quadratic = +6.00
- Cubic = -4.66
- Chi-Square = 0.68

Uncertainties:

- Shown for Pol3 Dpmjet, Pythia6, Pythia8, Ope & Upc.

# Asymmetry based on Polynomial, Power and Exponential Functions



Files range: 0 and 25 (different parameters & Chi2)

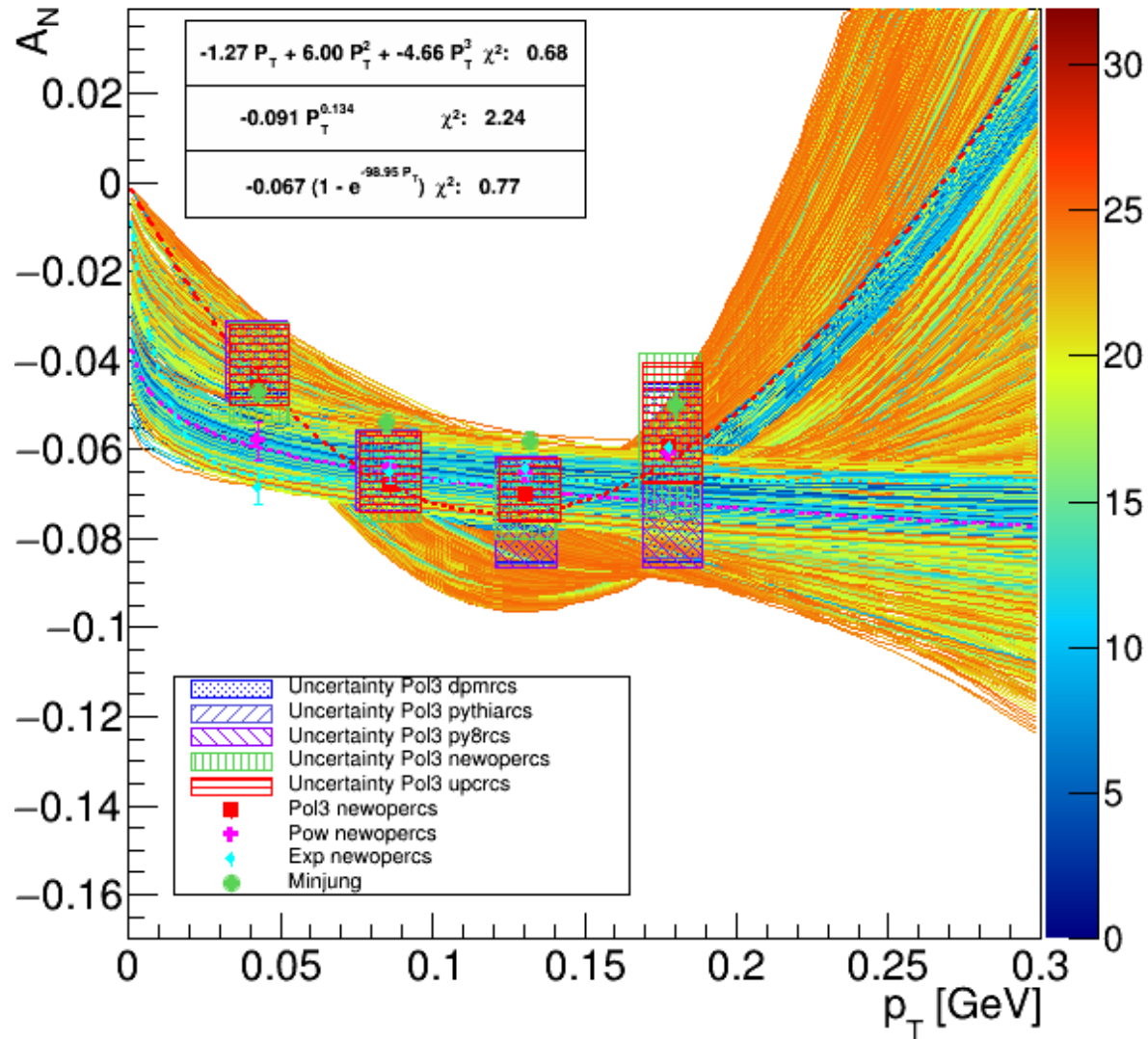
Best chosen parameters:

- Linear = -0.091
- Quadratic = +0.134
- Chi-Square = 2.24

Uncertainties:

- Shown for Pol3 Dpmjet, Pythia6, Pythia8, Ope & Upc.

# Asymmetry based on Polynomial, Power and Exponential Functions



Files range: 0 and 25 (different parameters & Chi2)

Best chosen parameters:

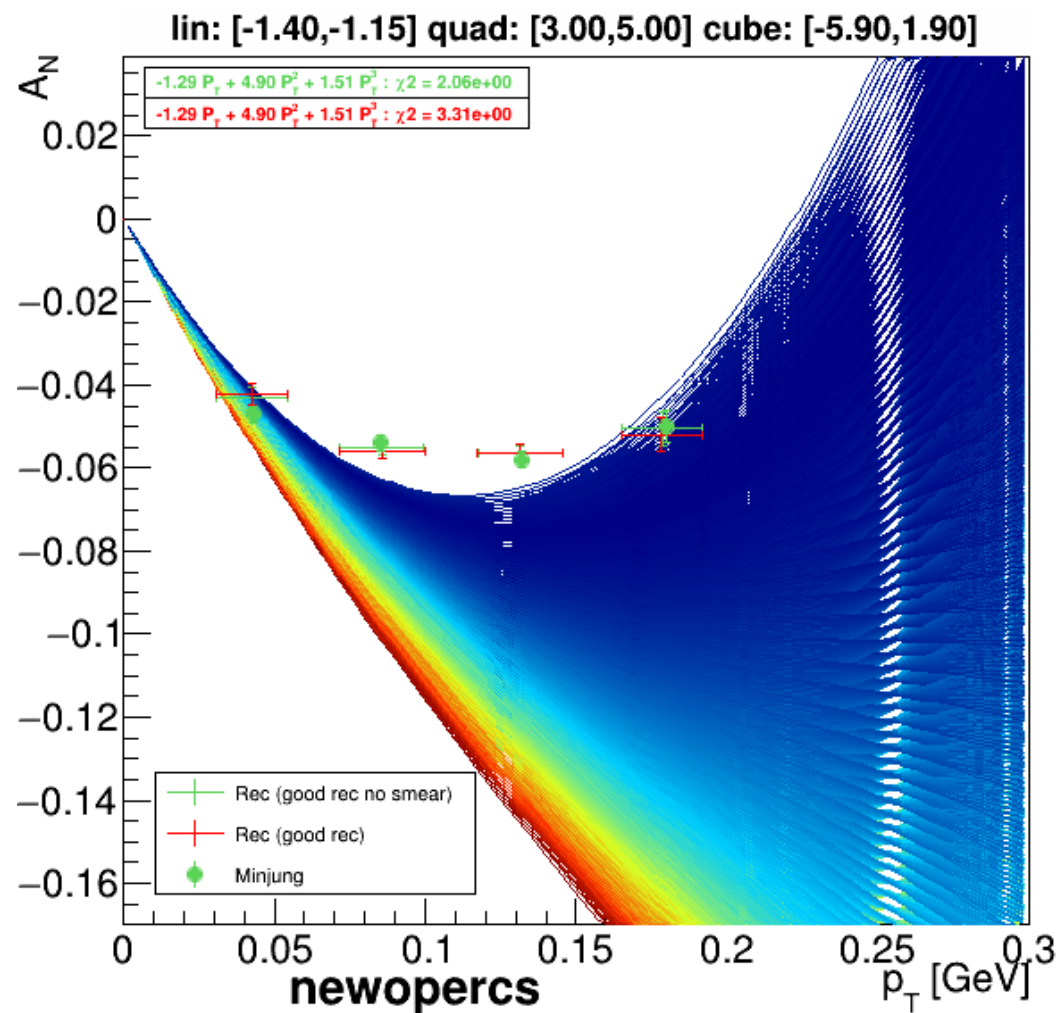
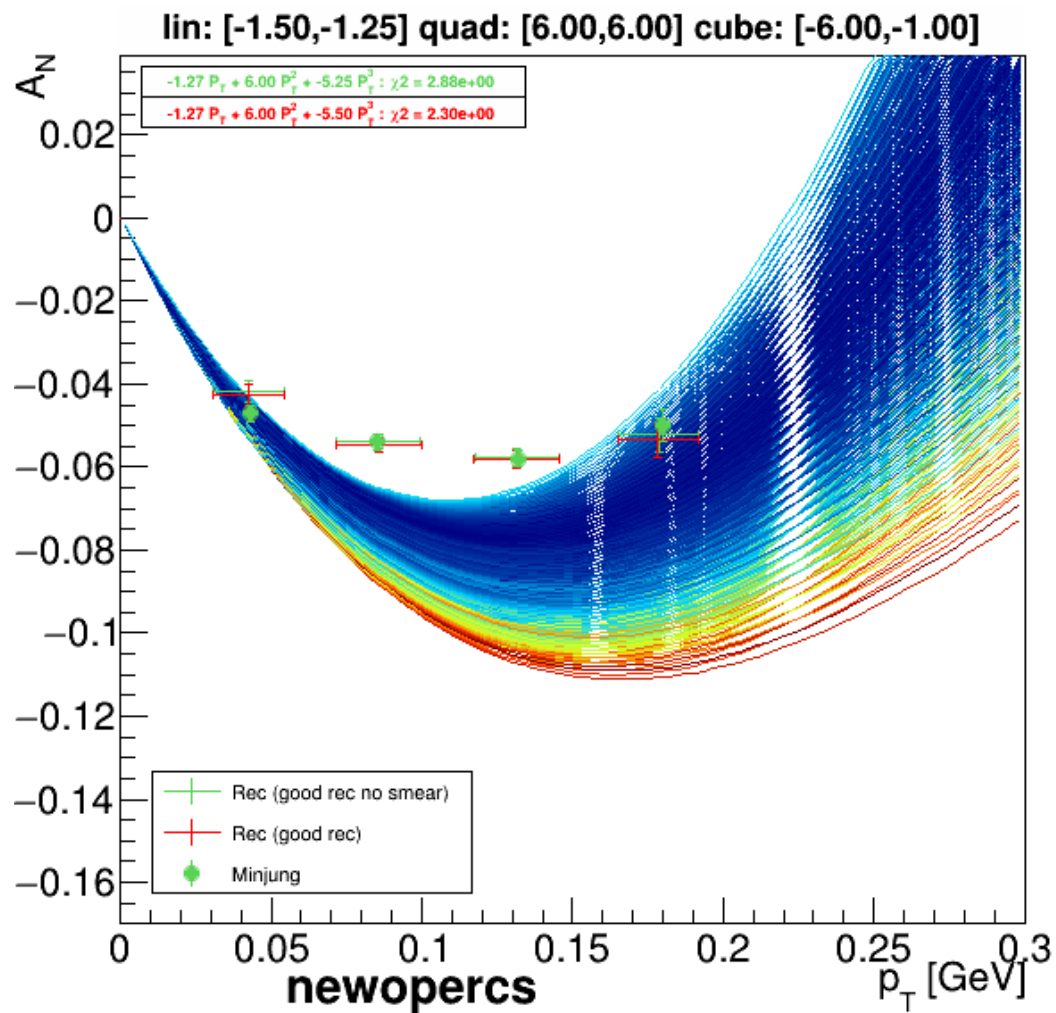
- Linear = -0.067
- Quad + Cube = 98.95
- Chi-Square = 0.77

Uncertainties:

- Shown for Dpmjet, Pythia6, Pythia8, Ope & Upc.



# Scan of Linear, Quadratic and Cubic Parameter Values



Need to zoom in by scanning a finer range of parameters for linear, quadratic and cubic and unfold

## Algorithm

1. Create two spin states using TRandom Number Generator:  
Spin up (0)  
Spin down (1)
2. Create spin depended weight according to Taylor series of a polynomial in the form:

$$w = 1 + (a + b * P_{T,T} + c * P_{T,T}^2 + d * P_{T,T}^3) \cos(\varphi_T + spin * \pi)$$

the parameters are:

**a** = constant

**b** = linear

**c** = quadratic

**d** = cube

spin \* pi = phase shift

spin = 0 (up)

1 (down)

**Note:** Other functional forms can also be scanned and tried to describe data asymmetries.

## Algorithm...

3. Scan parameters for different functional forms over a wide range using chi-square based on the reconstructed asymmetries from pp collision monte carlo samples and run 15 pp asymmetry results (Minjung's result) to find the best parameter, i.e. parameter with lowest,

$$\chi^2 = \sum_i \frac{(A_{N,i}^{Minjung} - A_{N,i}^{w,reco})^2}{(\Delta A_{N,i}^{2,Minjung} + \Delta A_{N,i}^{2,w,reco})}$$

4. Extract the asymmetry using the best Chi-squared parameters,

$$A_N = \frac{N_{\Phi\uparrow} - N_{\Phi\downarrow}}{N_{\Phi\uparrow} + N_{\Phi\downarrow}}$$