

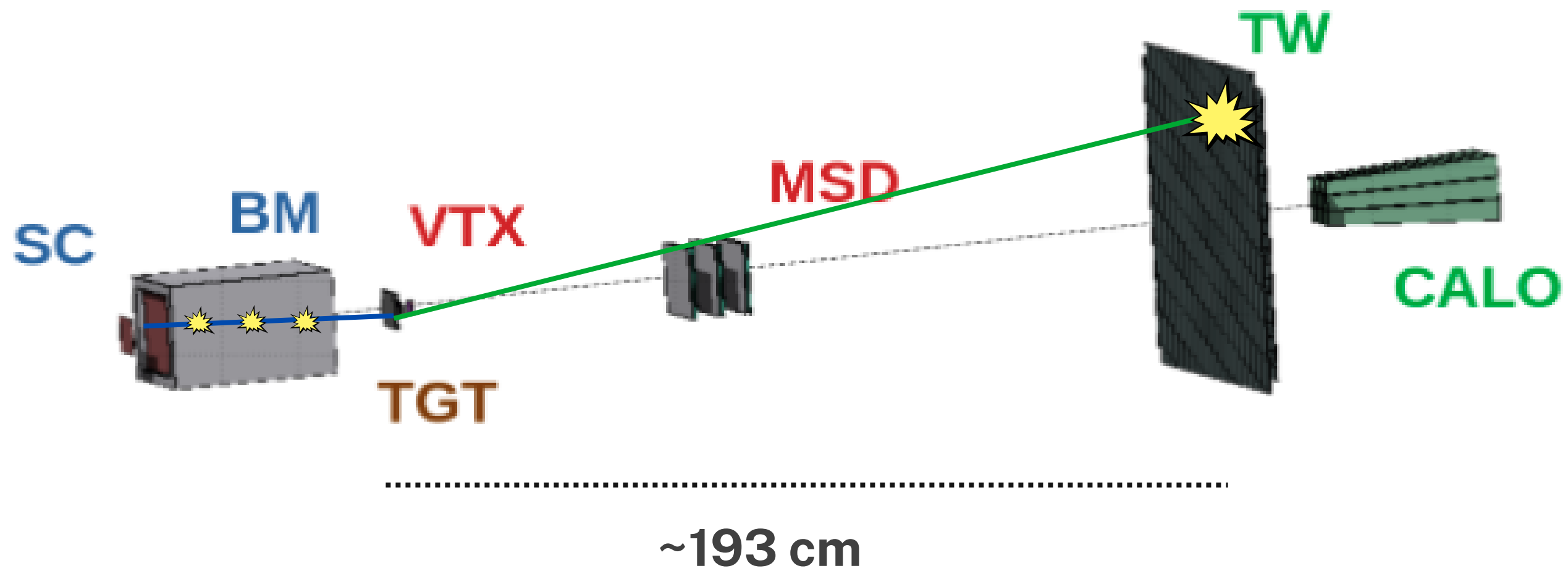


GSI2021 analysis without tracking

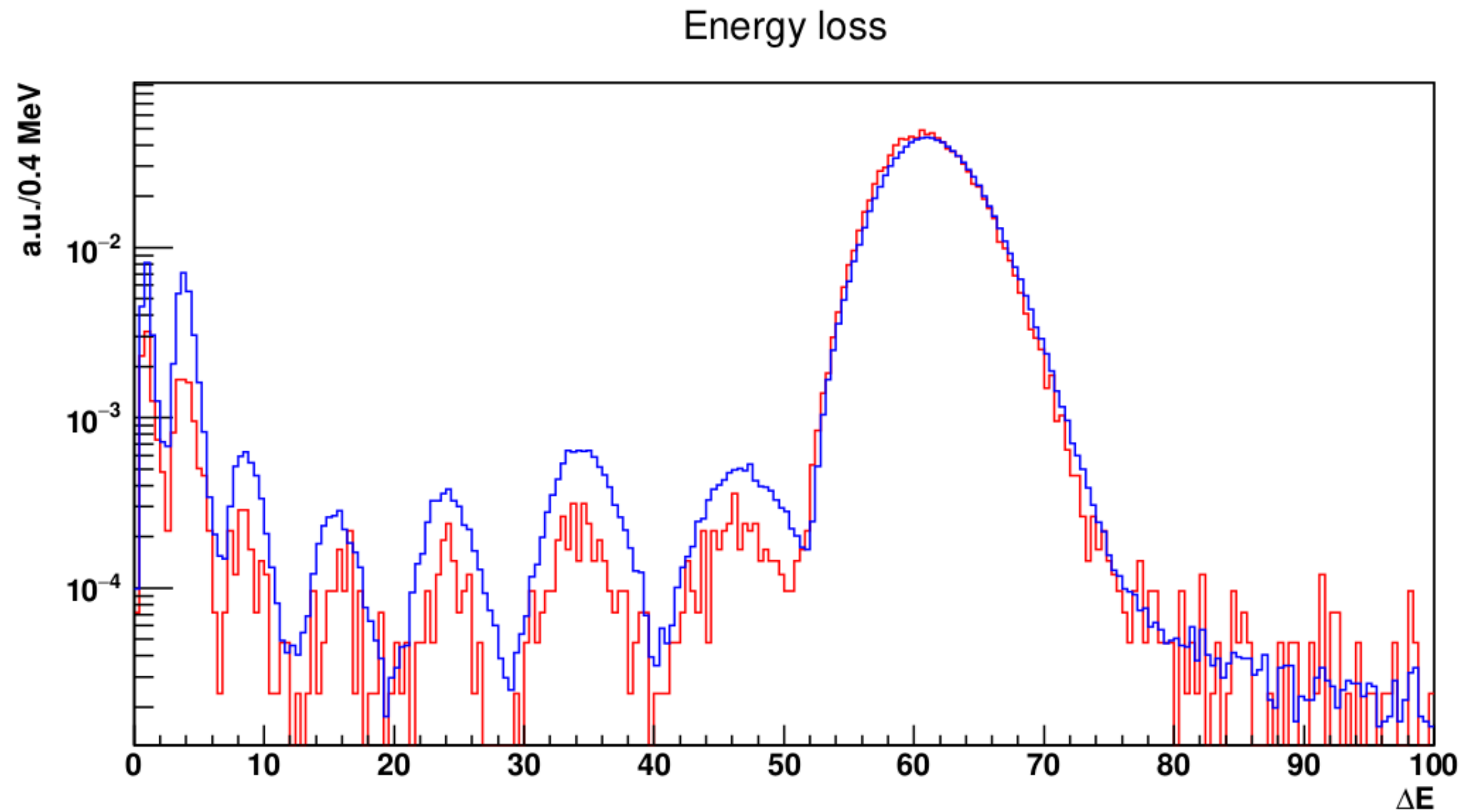
Riccardo Ridolfi

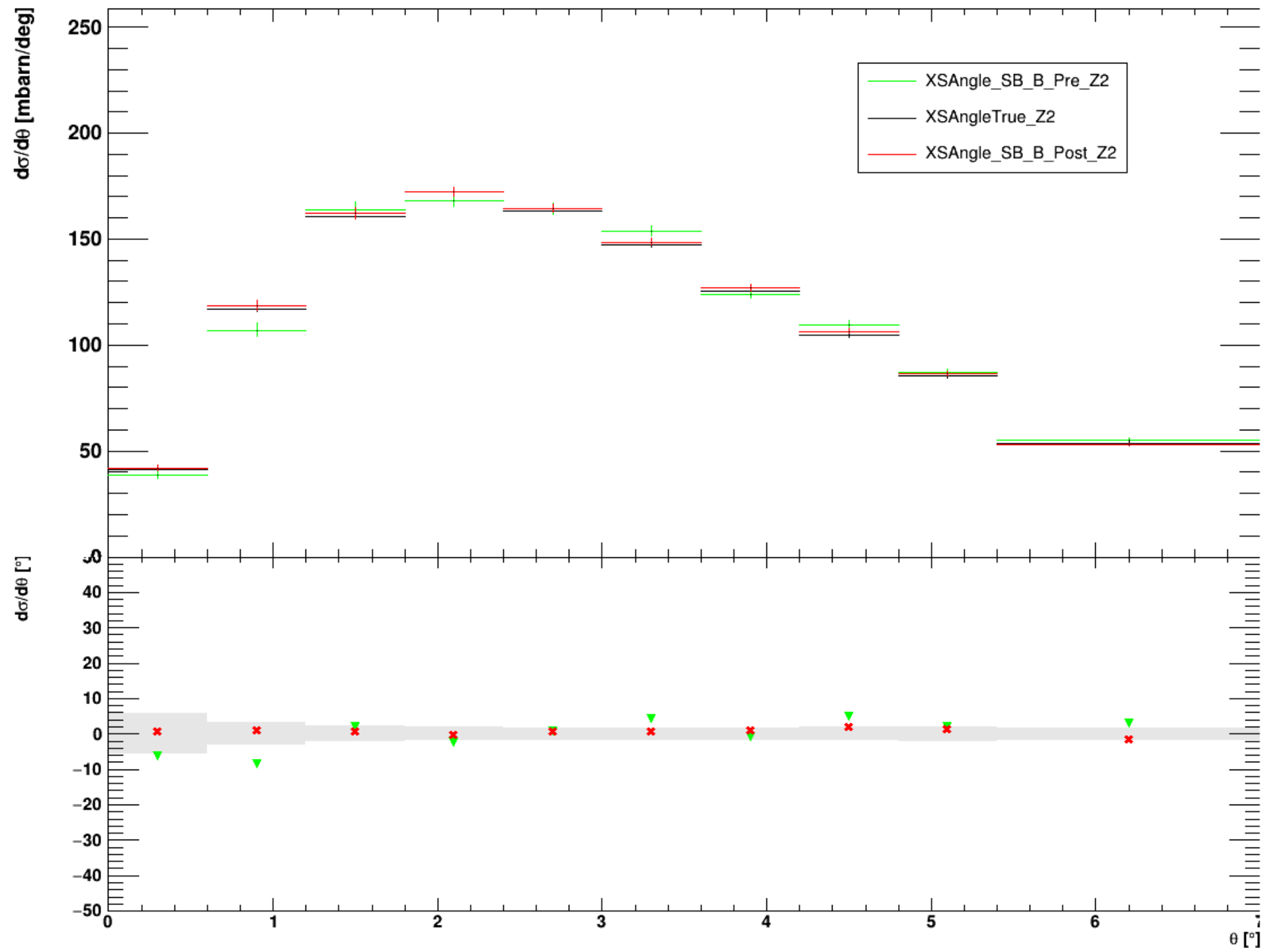
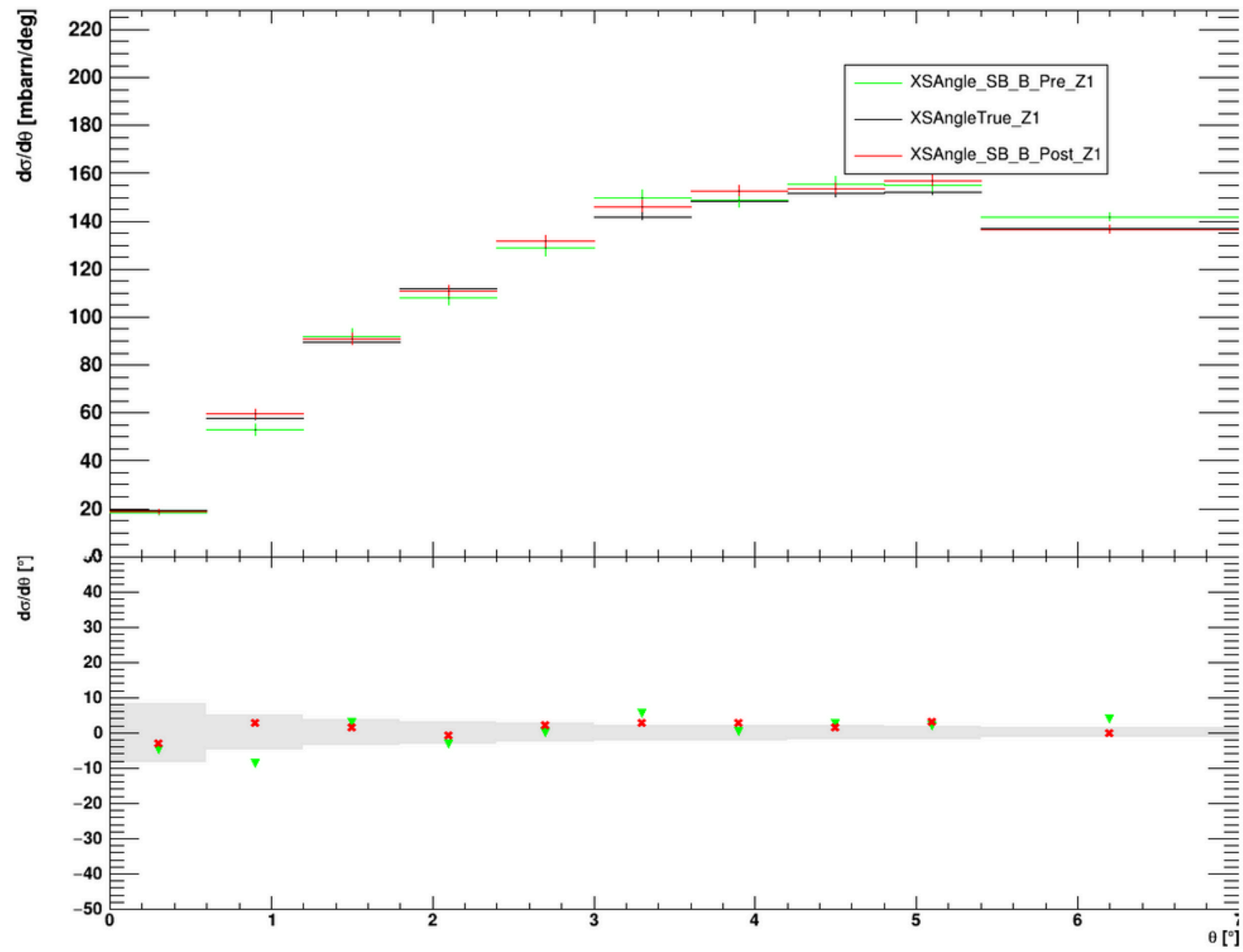
5 June 2024

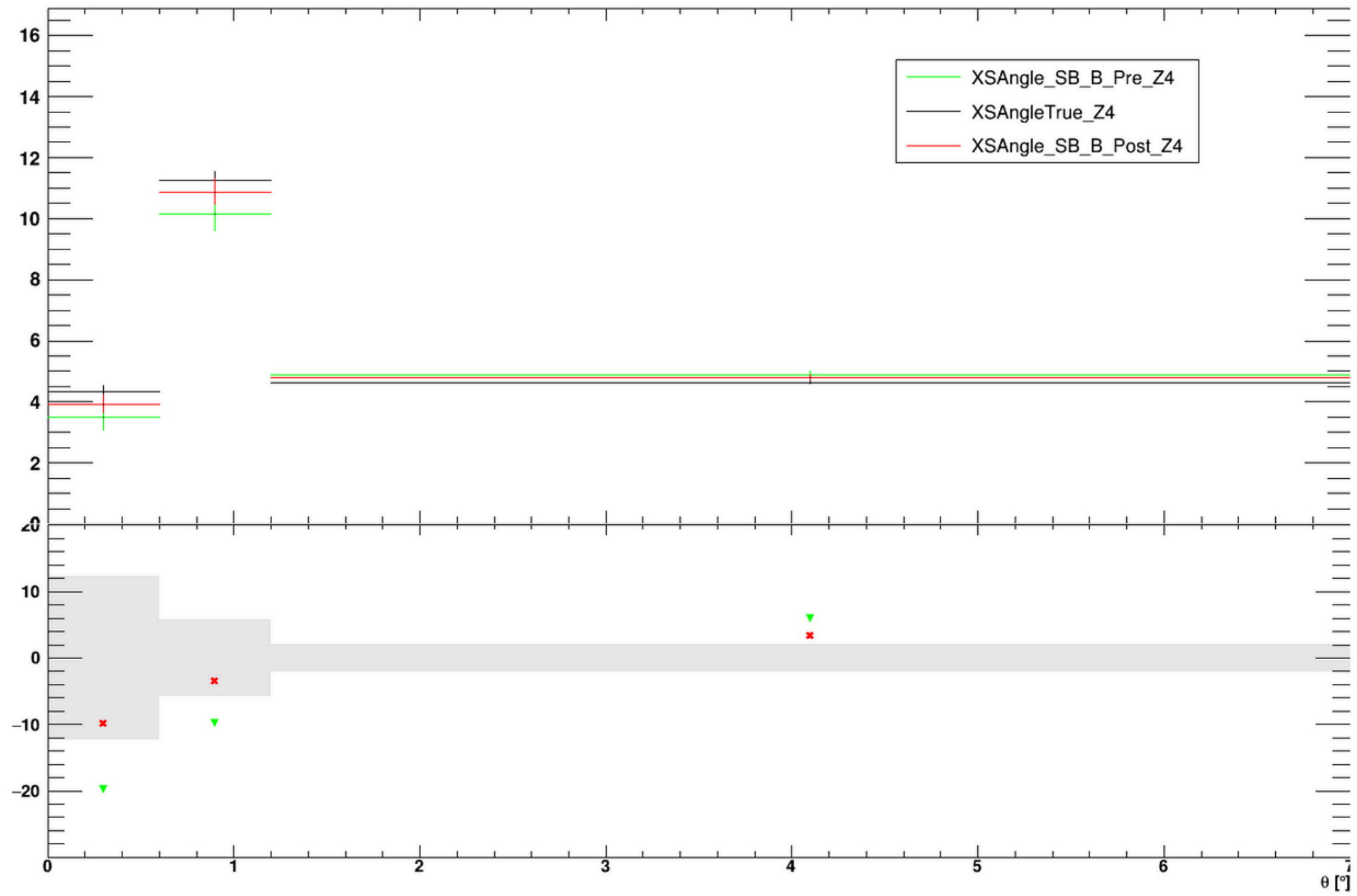
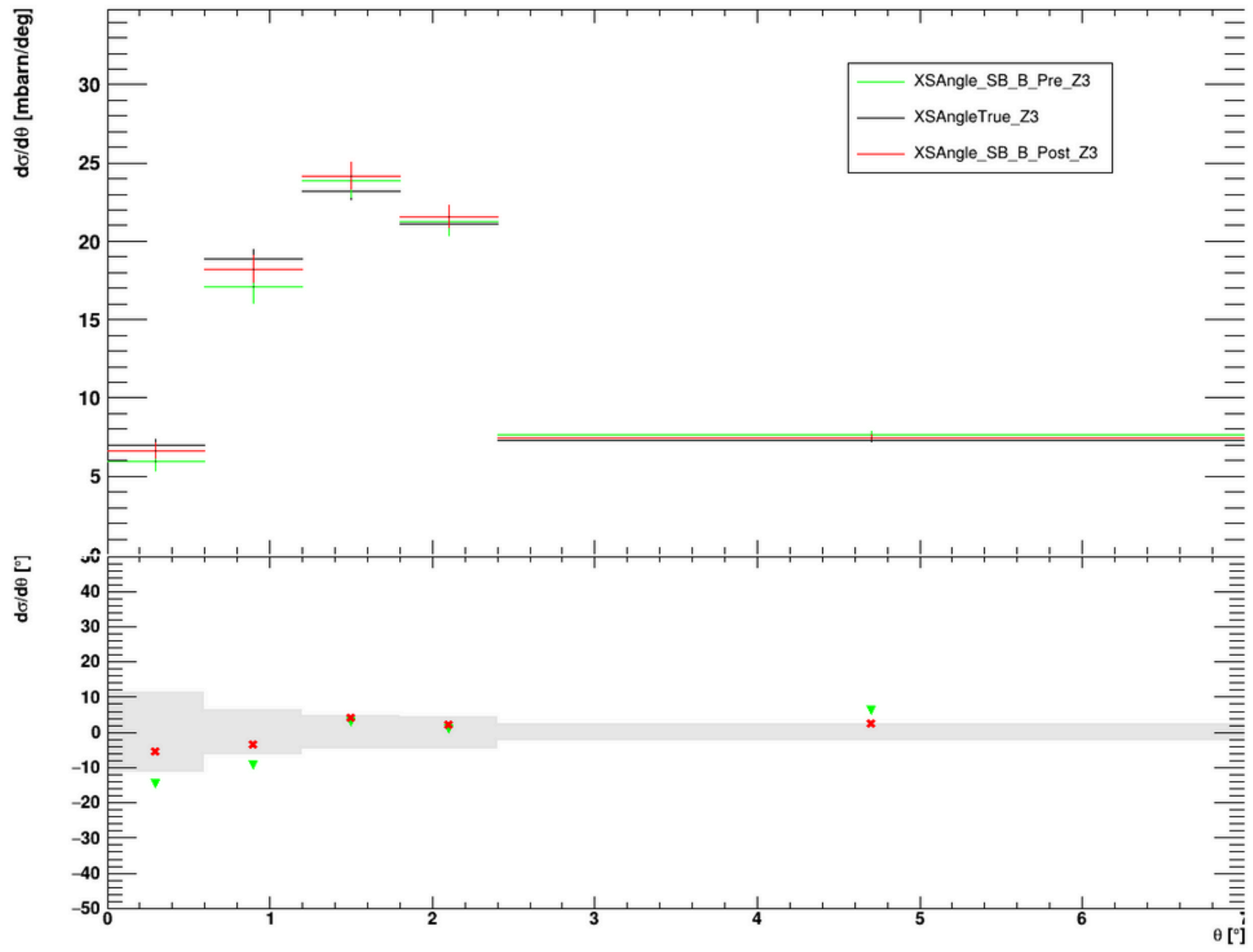
Angle measurement

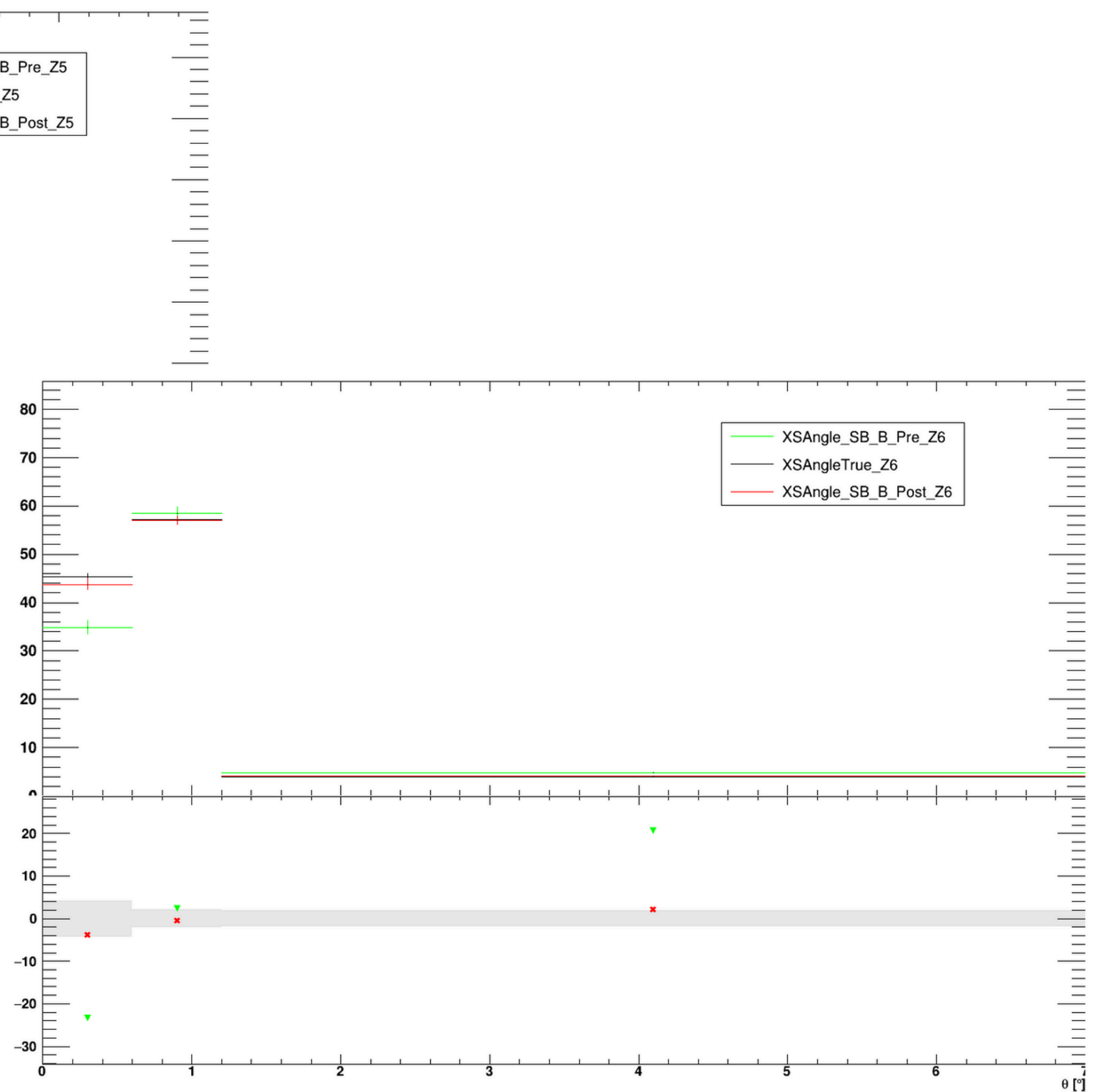
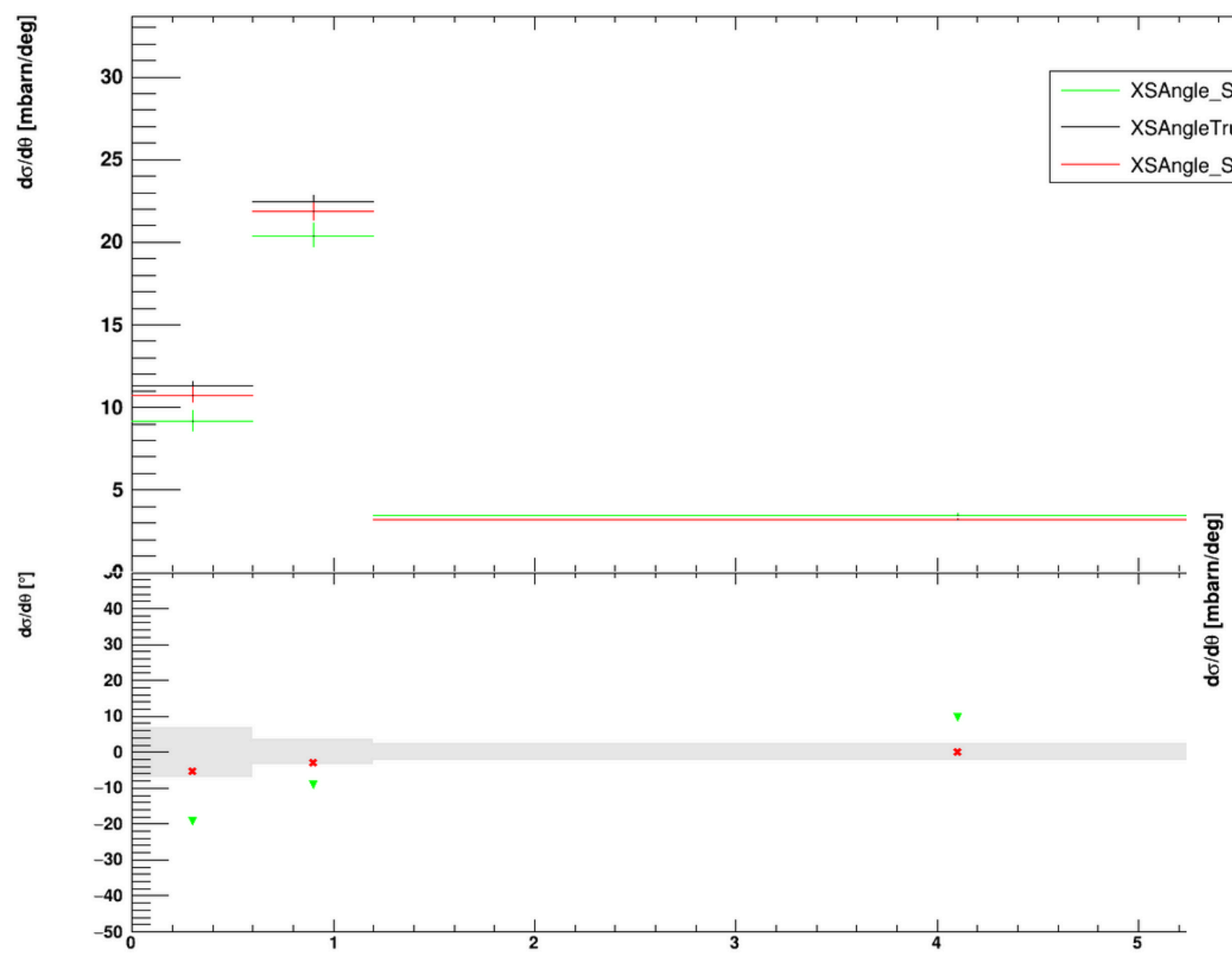


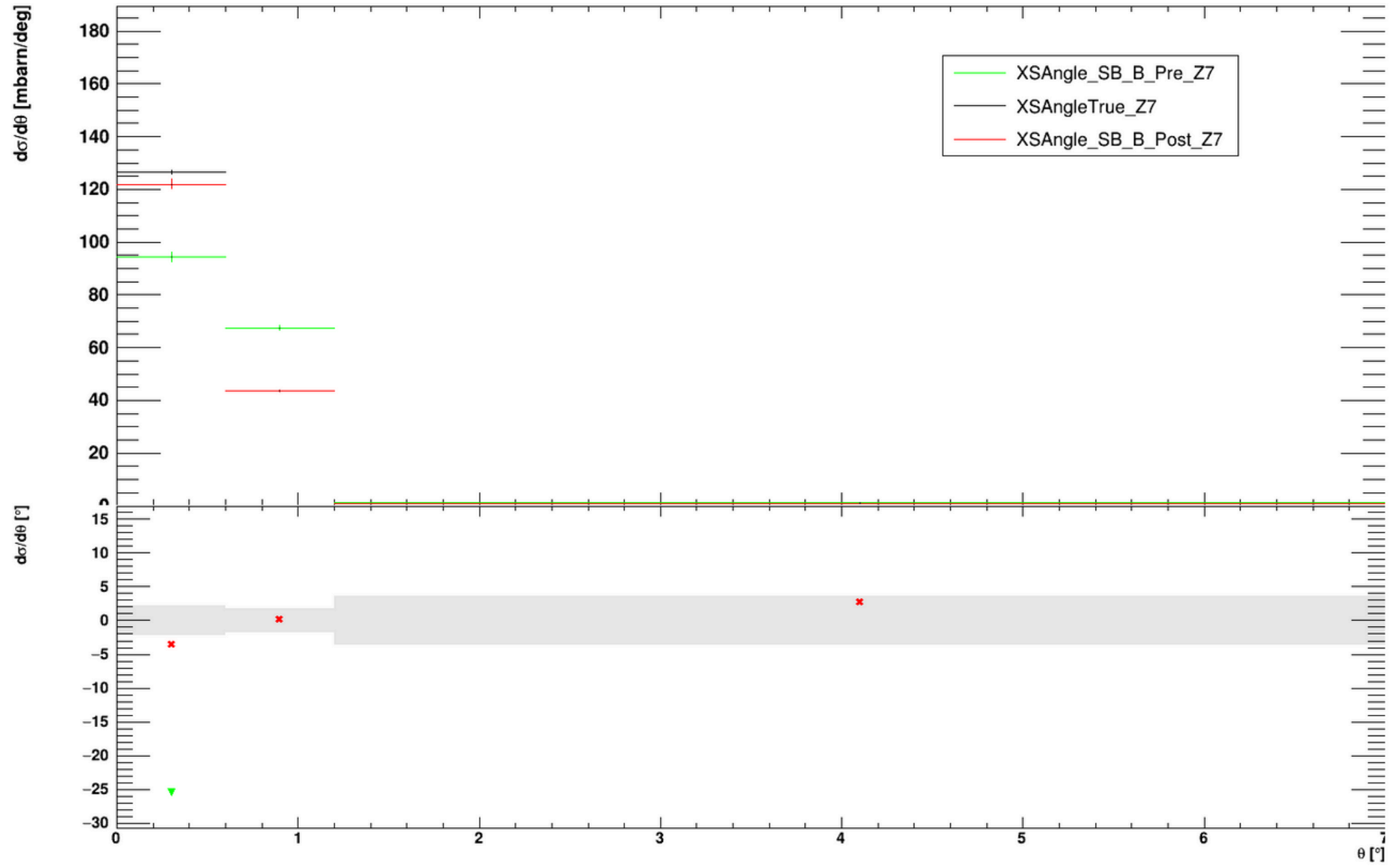
Why background subtraction?

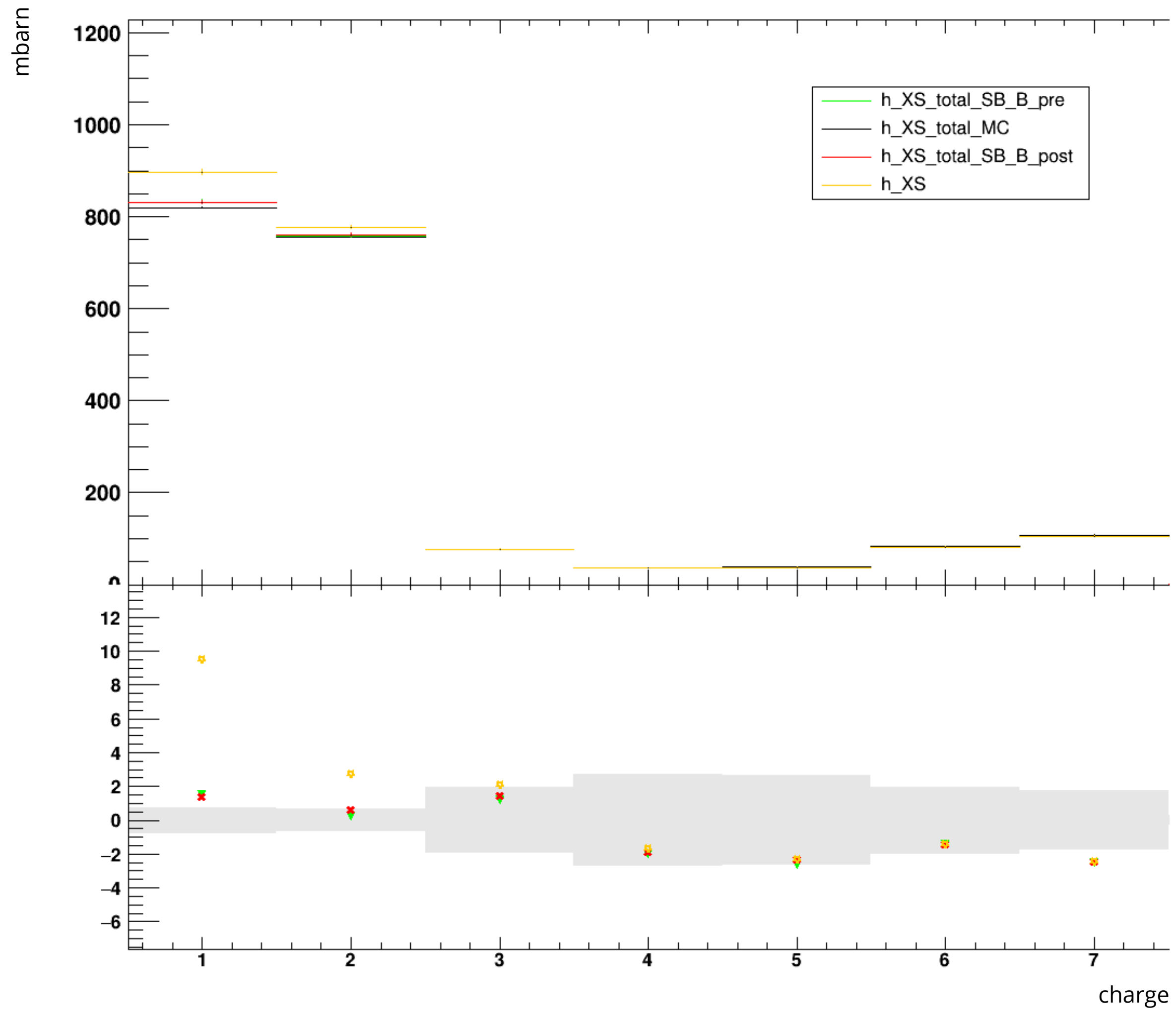










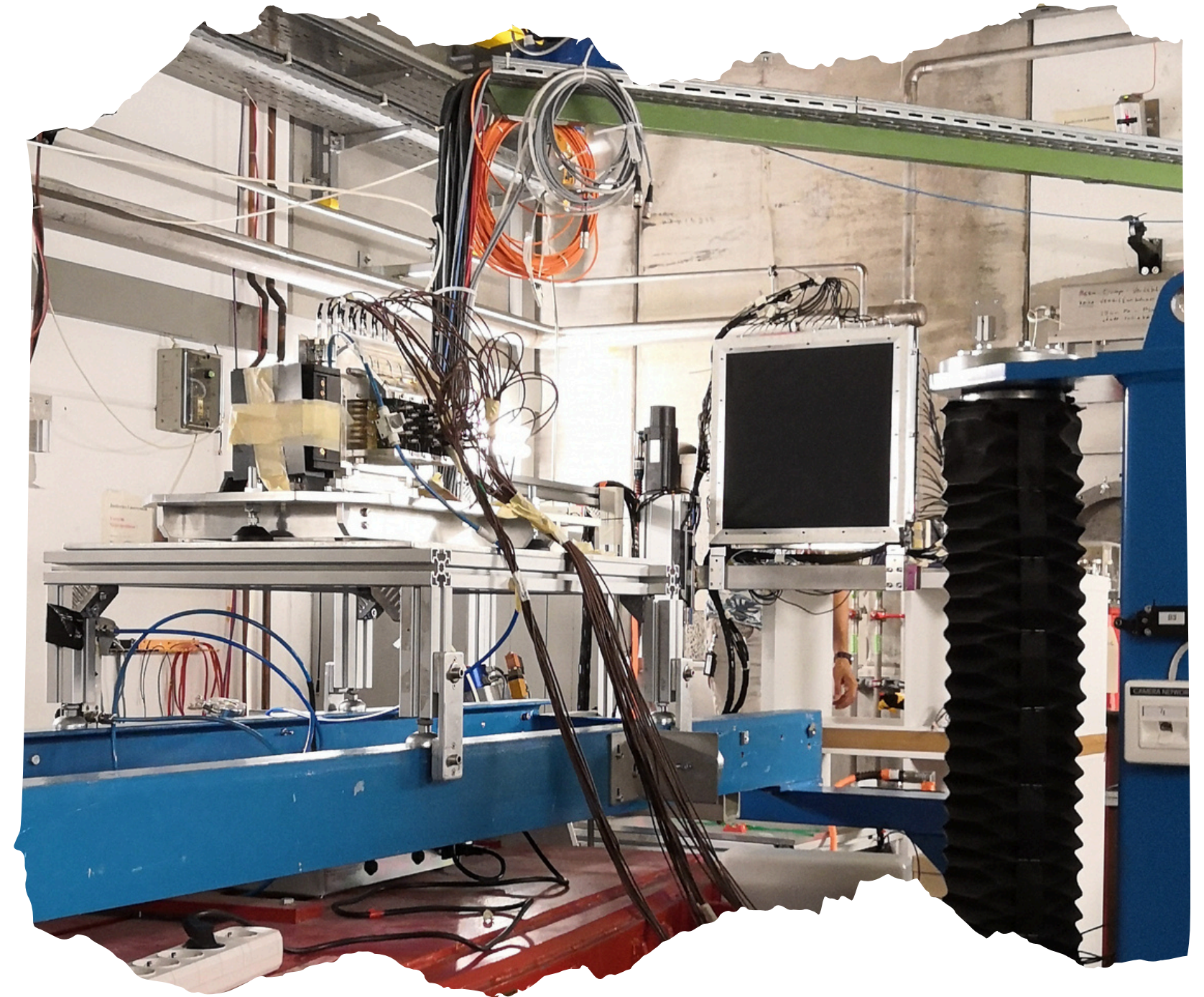


Data analysis

Run on data with the same steps of MC analysis

400 MeV/u ^{16}O beam on 5mm Carbon target

Run	Trigger type	Target	Events
4305	MB	C	162102
4306	MB	C	577096
4307	MB	C	513370
4308	Frag + MB	C	510169
4309	Frag + MB	C	531812
4310	Frag + MB	C	1012099
4313	MB	no	57133



New analysis flow

Evaluate efficiencies and purities

Repeat for with and w/o target samples

Apply reconstruction cuts (SC, BM)

Normalize yields and subtract background

Apply efficiency and purity for fragmentation in target

Unfolding

Calculate angular cross sections

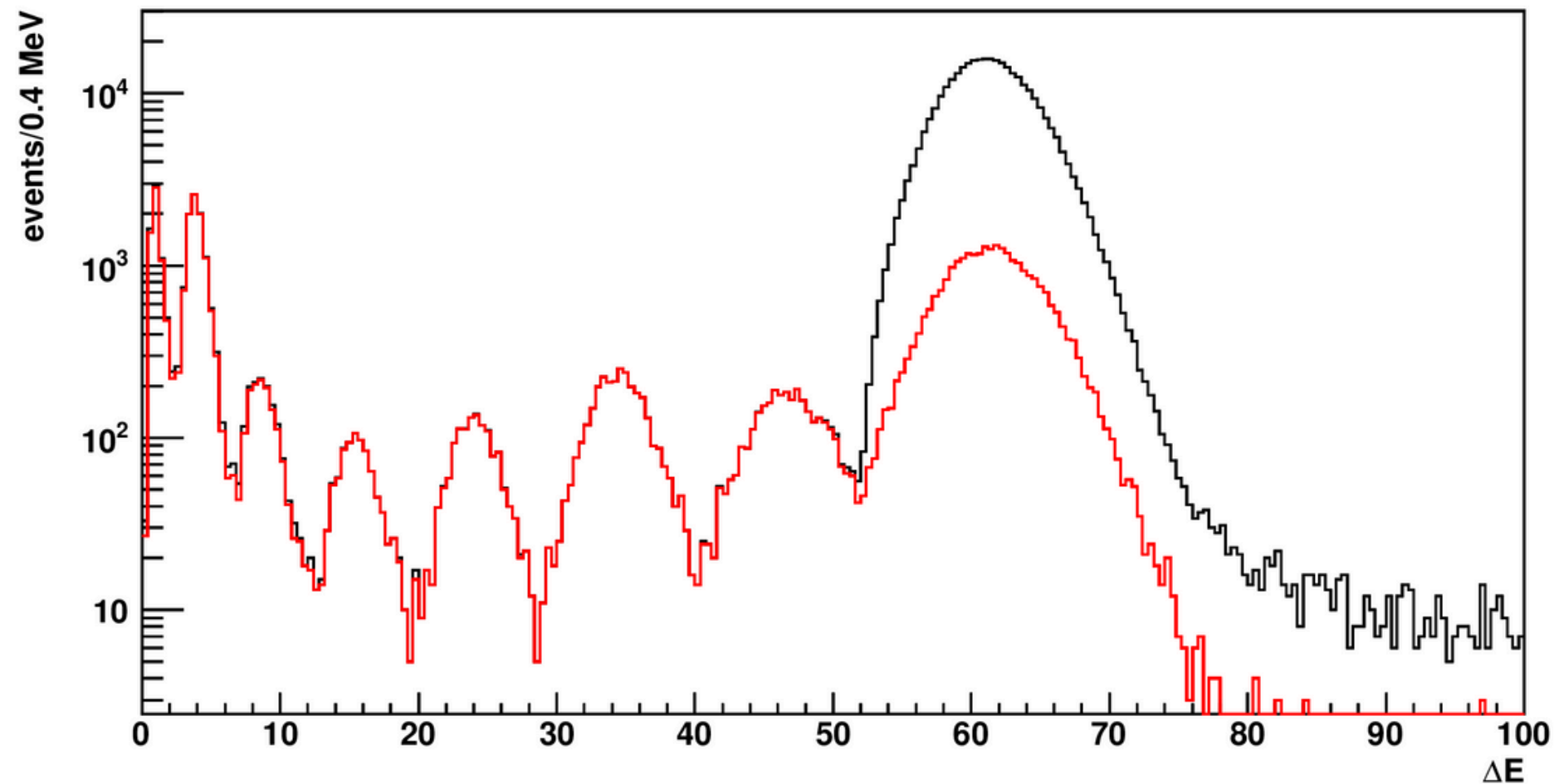
Add systematics uncertainties

Data analysis

In MB runs the number of primaries is the number of events passing selection cuts

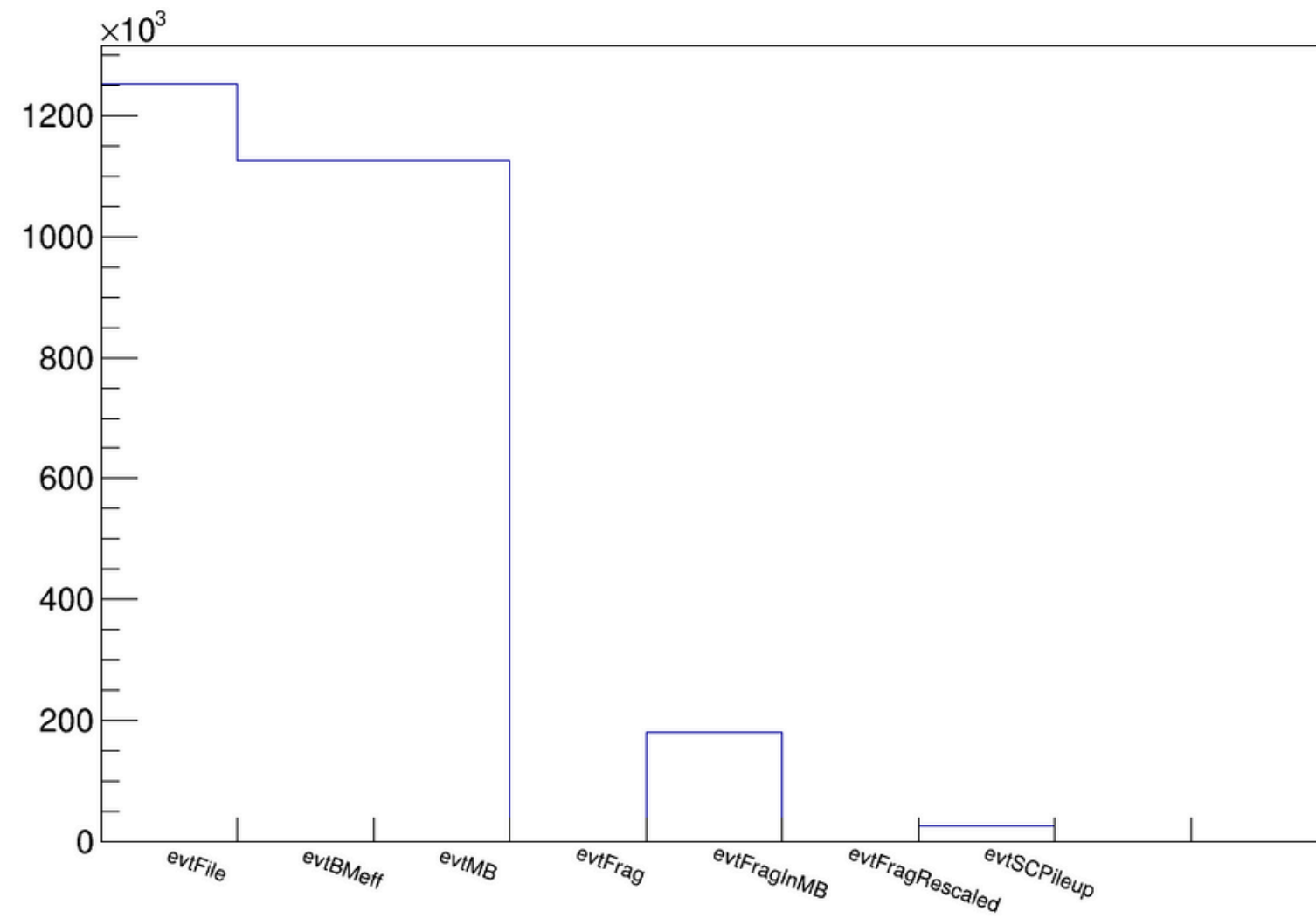
In fragmentation runs the number of primaries has to take into account the trigger rejection factor

It can be evaluated from MB runs (fragmentation flag: ON)

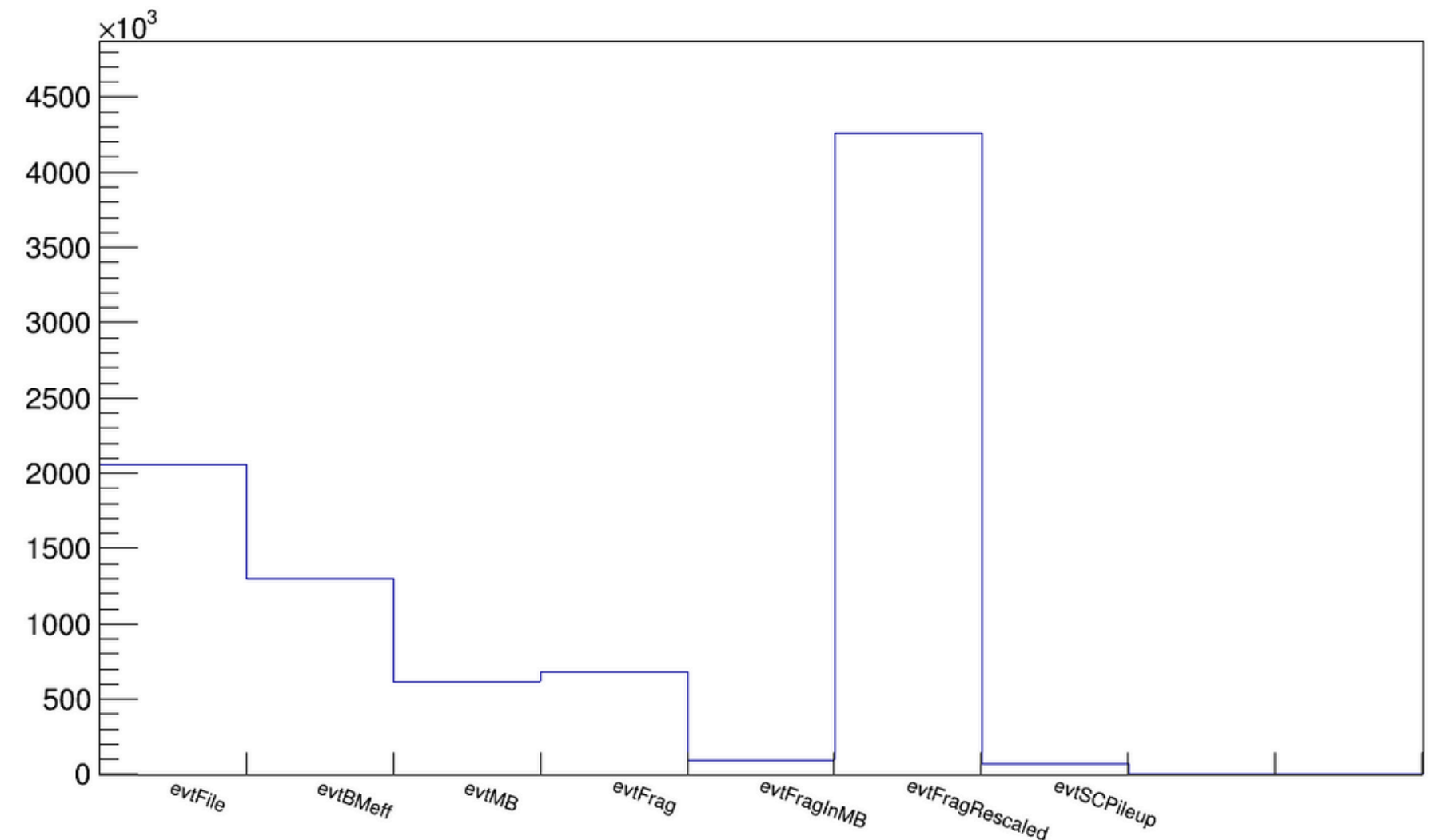


Number of events

Minimum bias (4305,4306,4307)

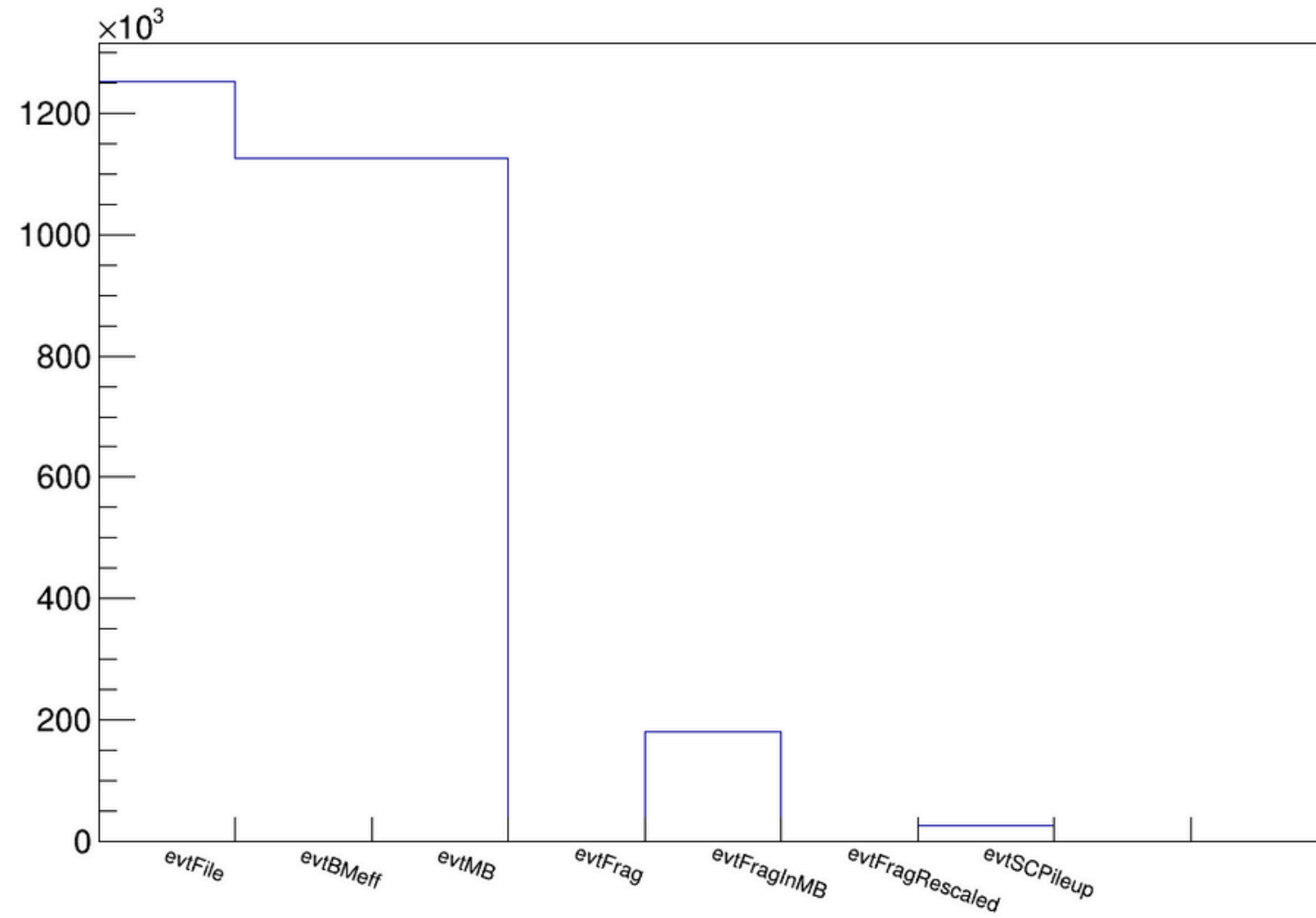


Fragmentation+MB (4308, 4309, 4310)

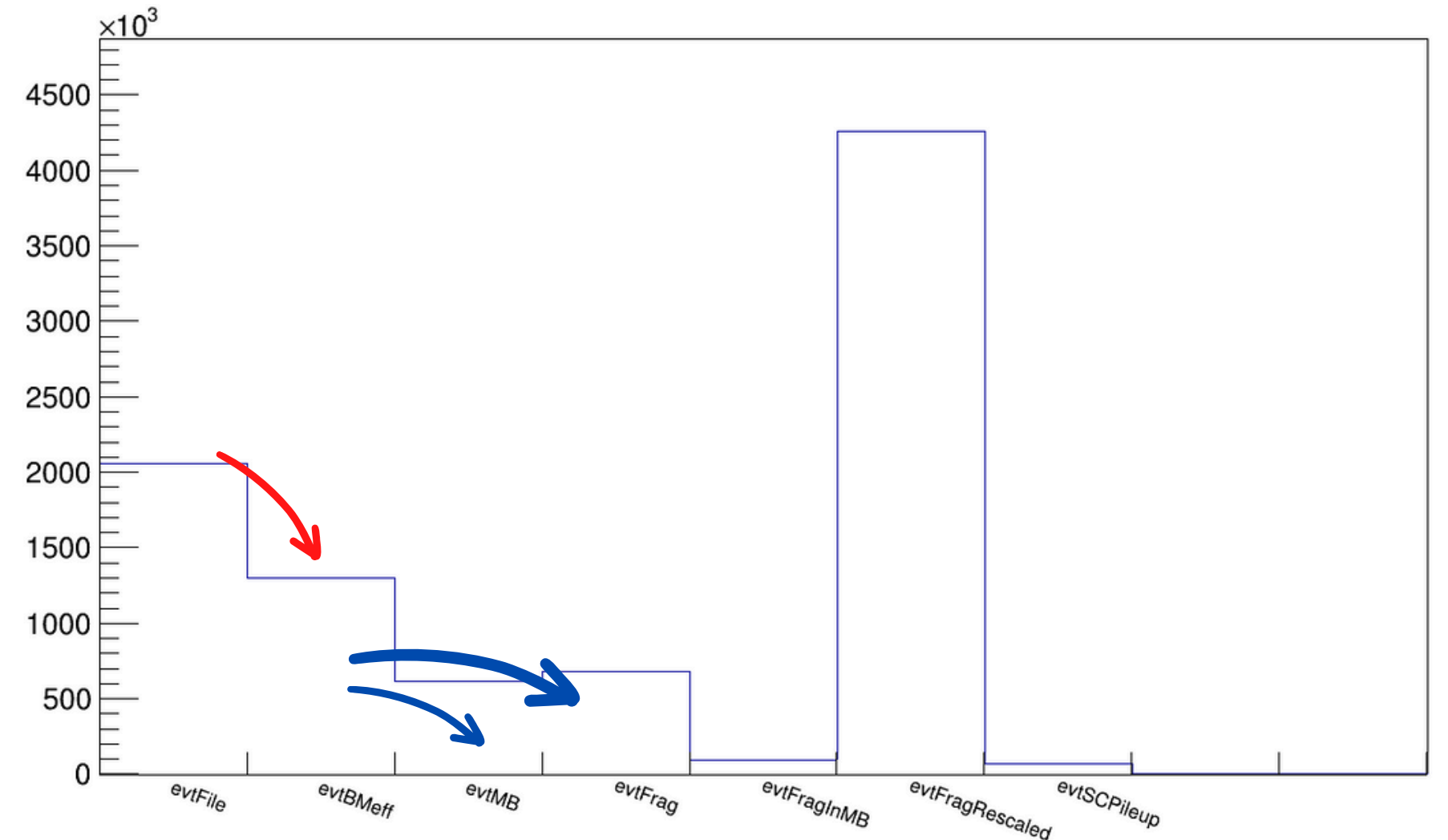


Number of events

Minimum bias (4305,4306,4307)

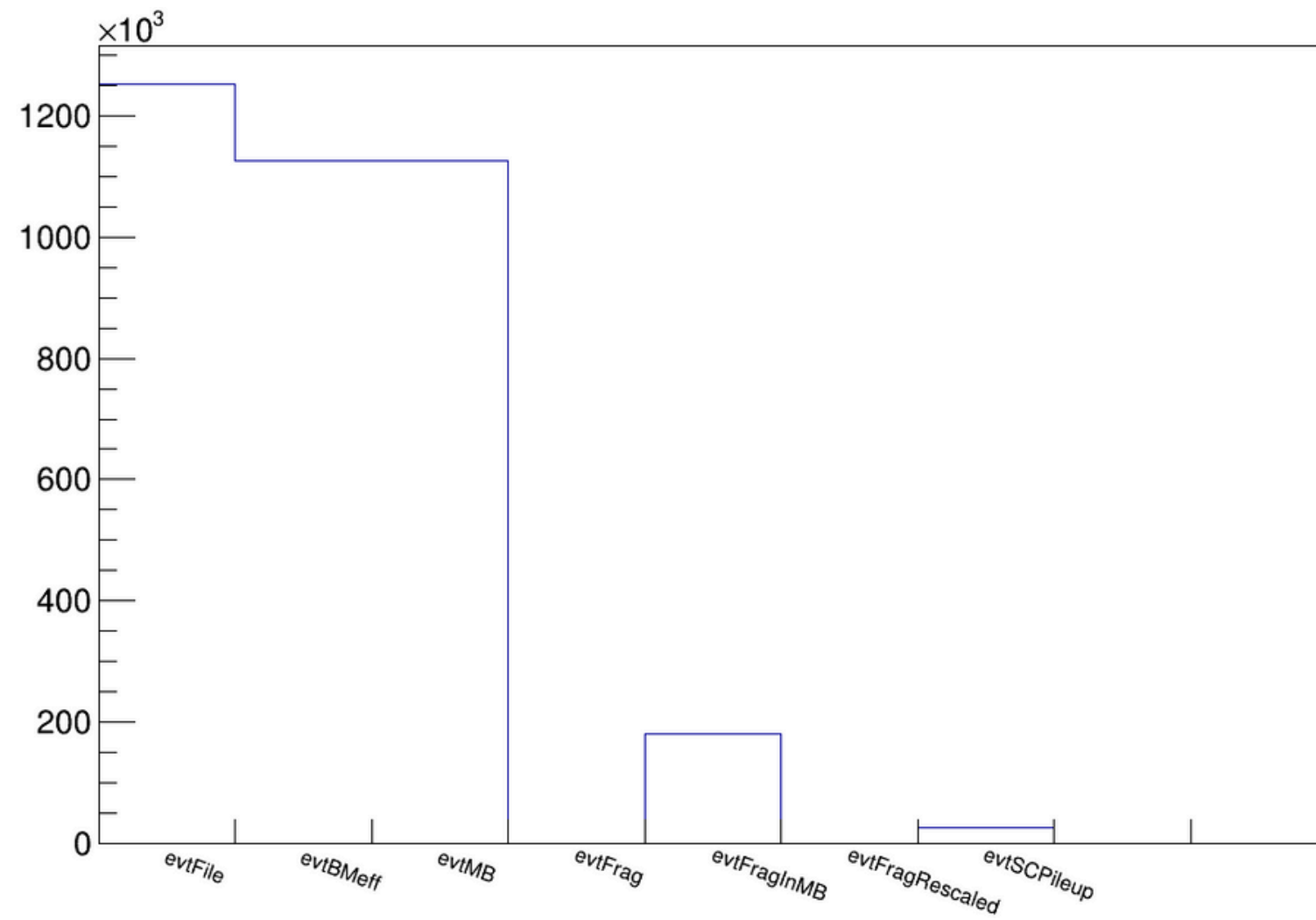


Fragmentation+MB (4308, 4309, 4310)

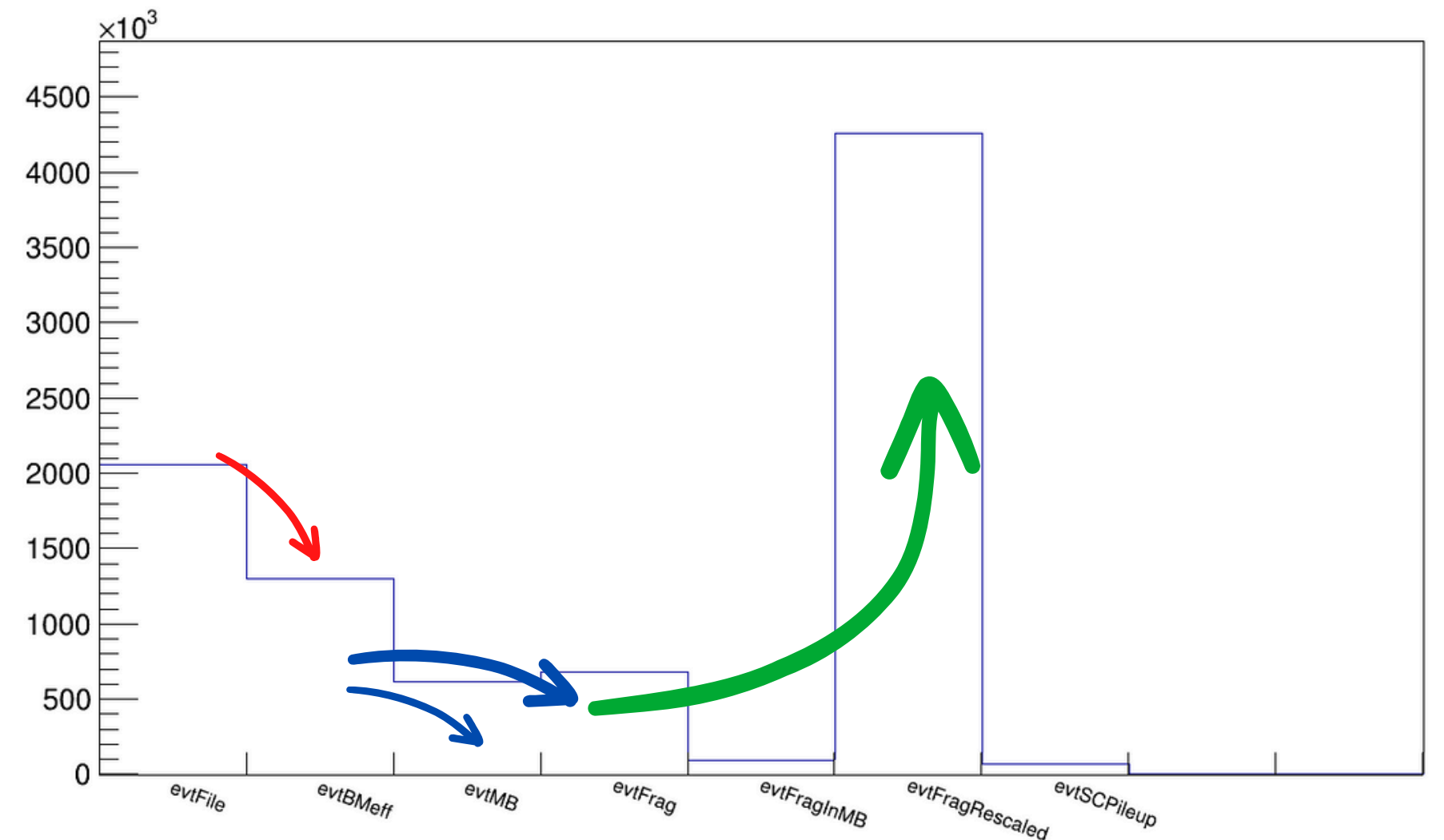


Number of events

Minimum bias (4305,4306,4307)



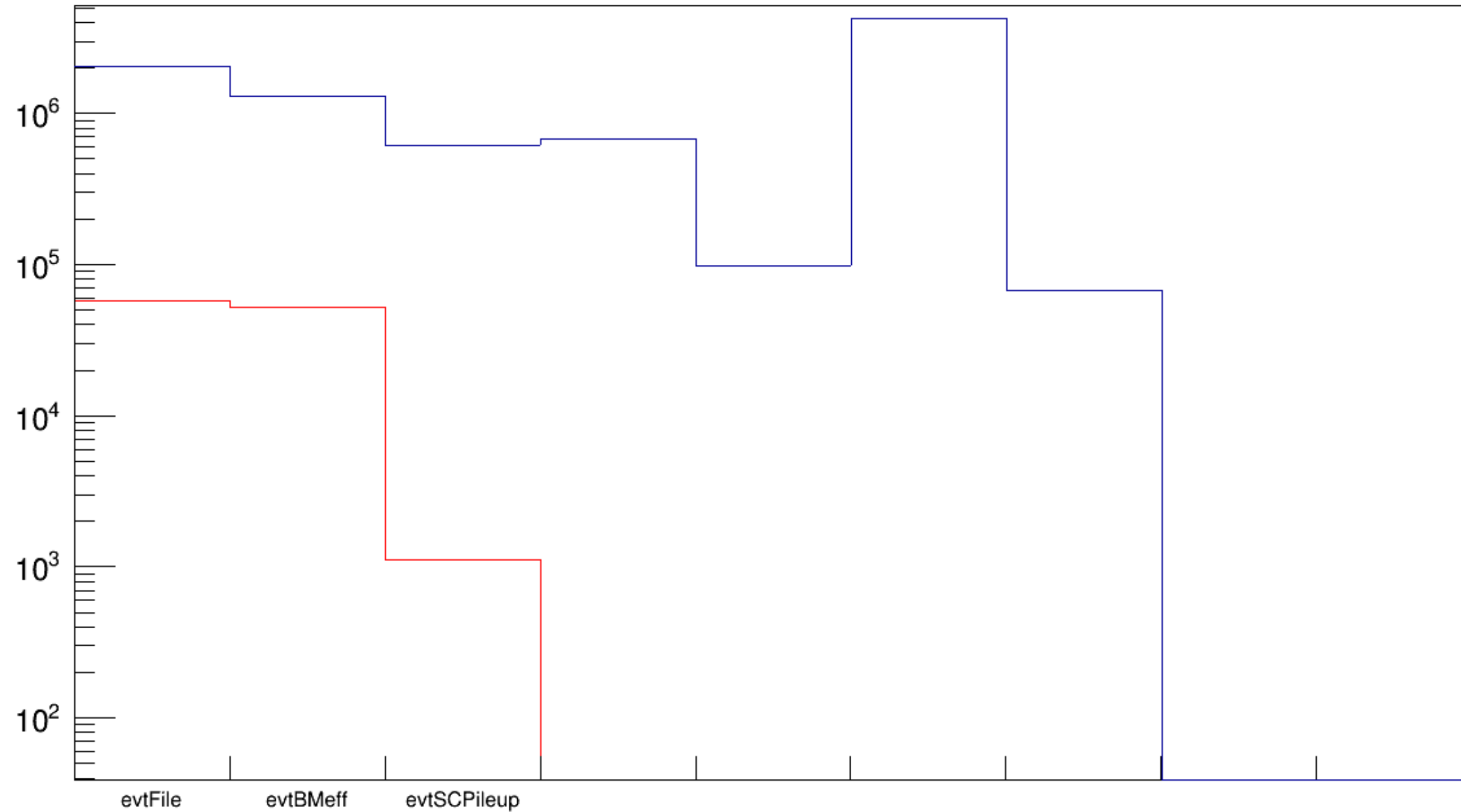
Fragmentation+MB (4308, 4309, 4310)



Number of events

Fragmentation+MB (4308, 4309, 4310)

Background MB (4313)



Impact of statistics on XS resolution

Relative uncertainties in XS (only stat)

$$\sigma(Z) = \frac{1}{N_{\text{TG}} \cdot \varepsilon(Z)} \cdot \left(\frac{Y_S(Z)}{N_S} - \frac{Y_B(Z)}{N_B} \right) = \frac{1}{N_{\text{TG}} \cdot \varepsilon(Z)} \cdot (S(Z) - B(Z))$$

$$\frac{\Delta\sigma}{\sigma} \approx \left(\frac{1}{S - B} \right) \cdot \sqrt{S^2 \cdot \left[\left(\frac{\Delta Y_S}{Y_S} \right)^2 + \left(\frac{\Delta N_S}{N_S} \right)^2 \right] + B^2 \cdot \left[\left(\frac{\Delta Y_B}{Y_B} \right)^2 + \left(\frac{\Delta N_B}{N_B} \right)^2 \right]}$$

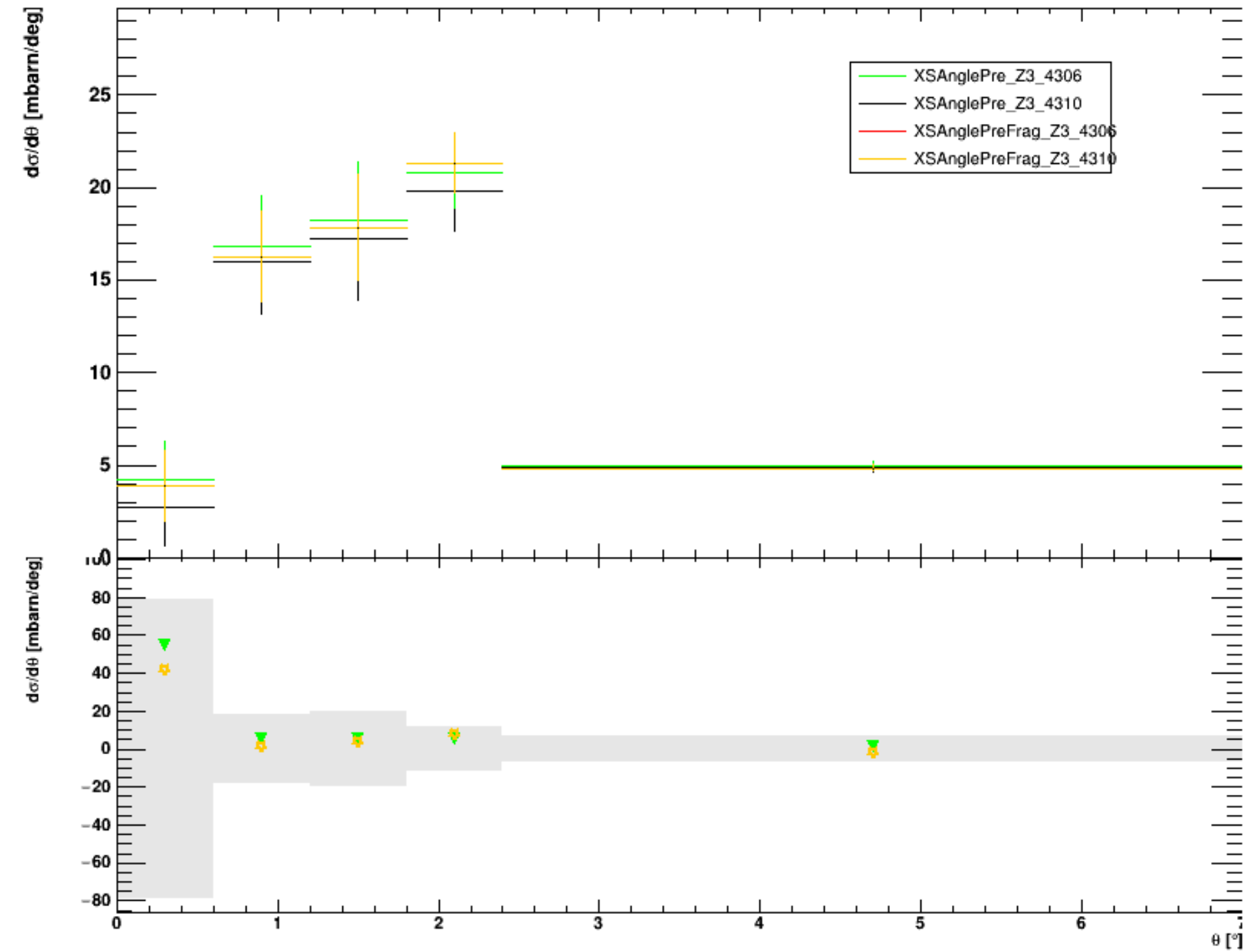
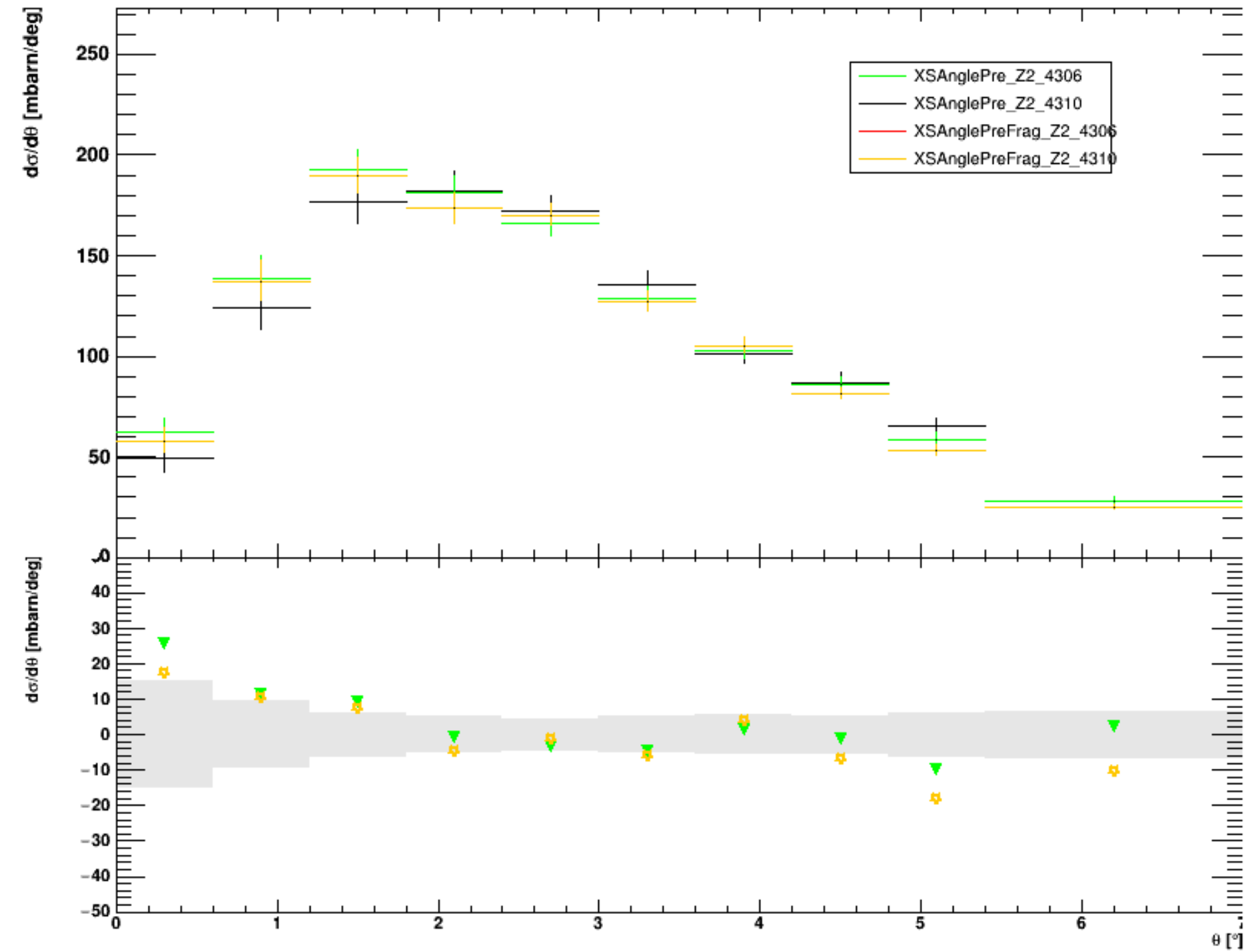
Fragmentation physics

Available Statistics

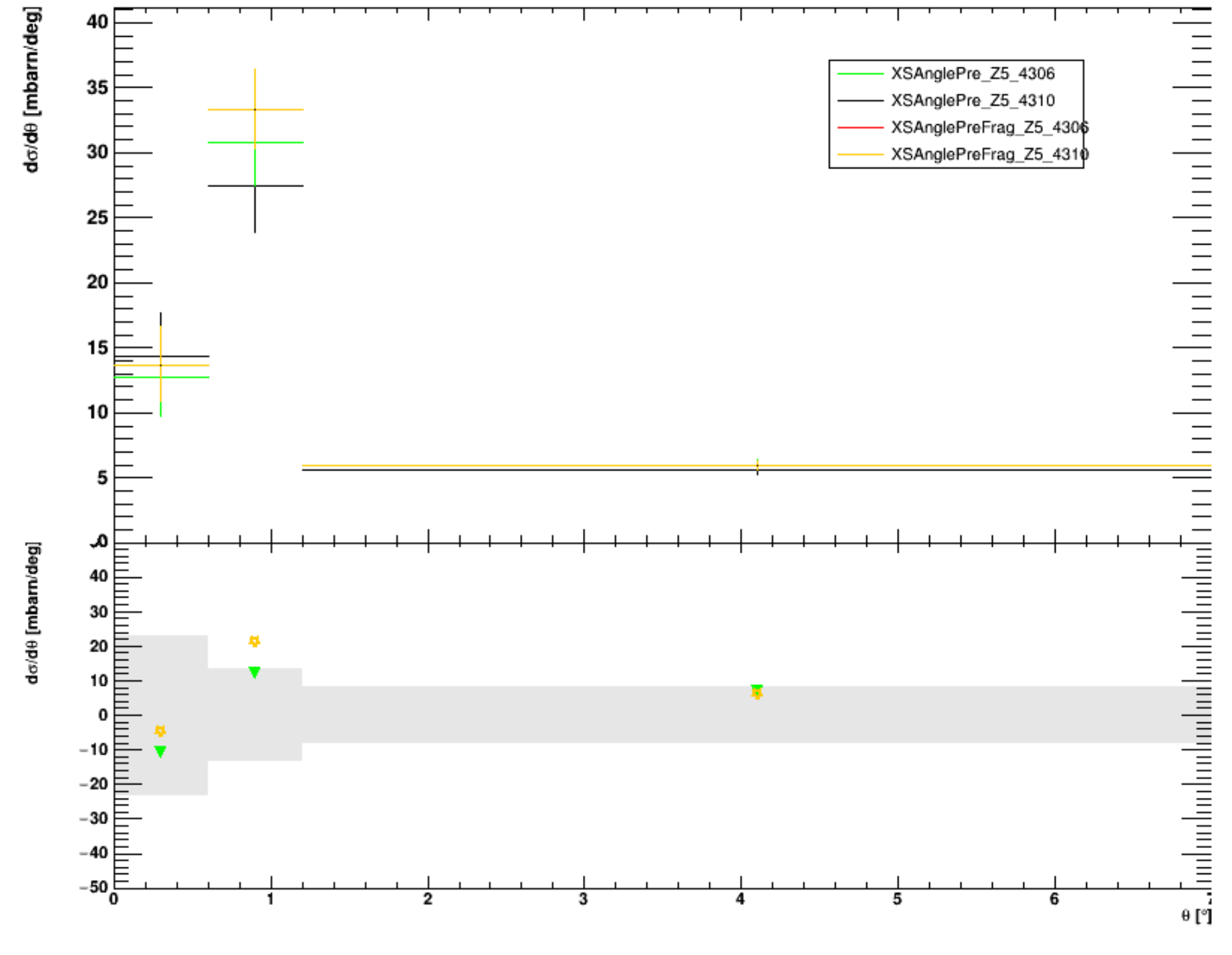
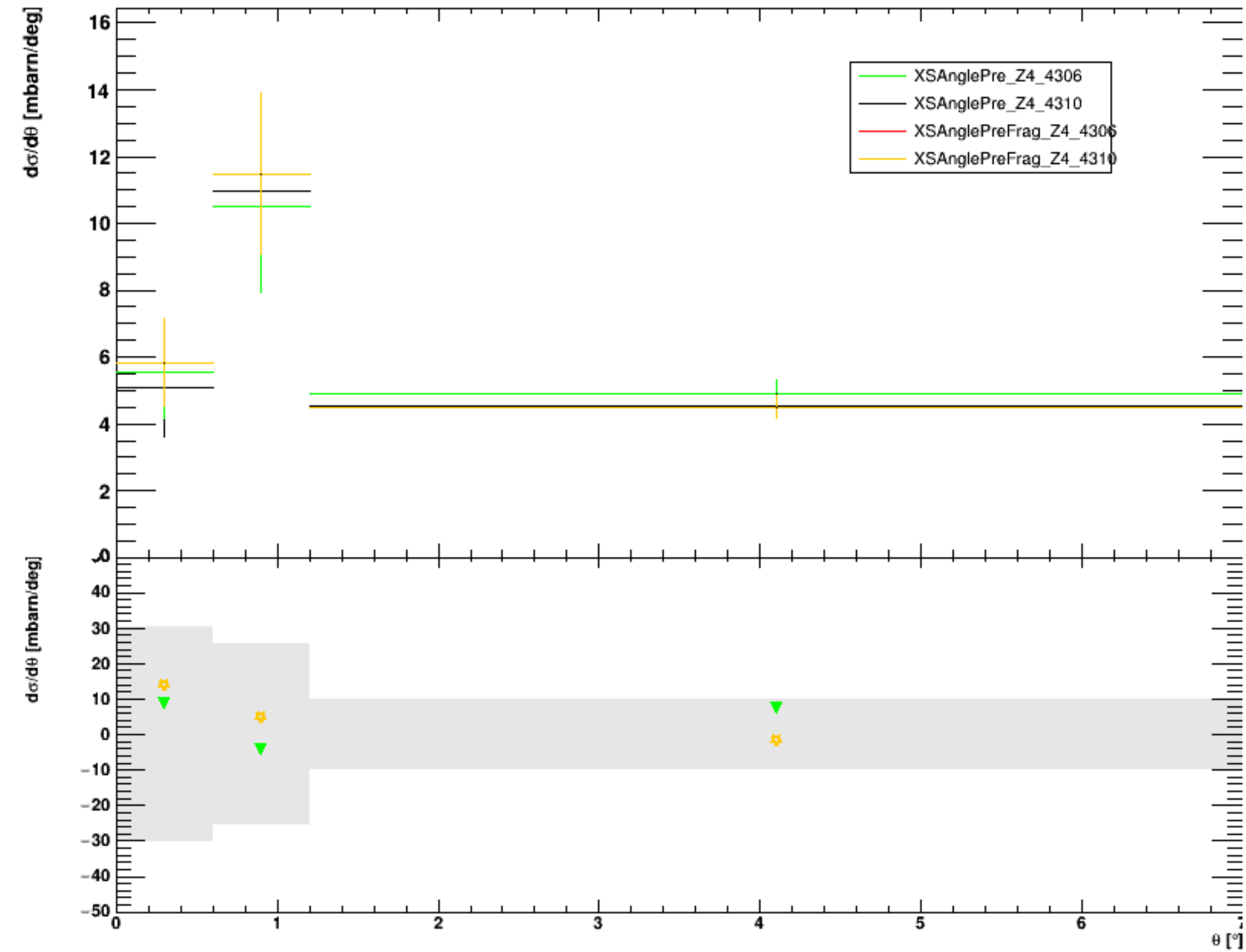
$$S = \frac{Y_S}{N_S} \quad B = \frac{Y_B}{N_B}$$

- Y_S fragments yields in TG runs (S->S+B)
- N_S primaries in TG runs (S->S+B)
- Y_B fragments yields in NO TG runs
- N_B primaries in NO TG runs

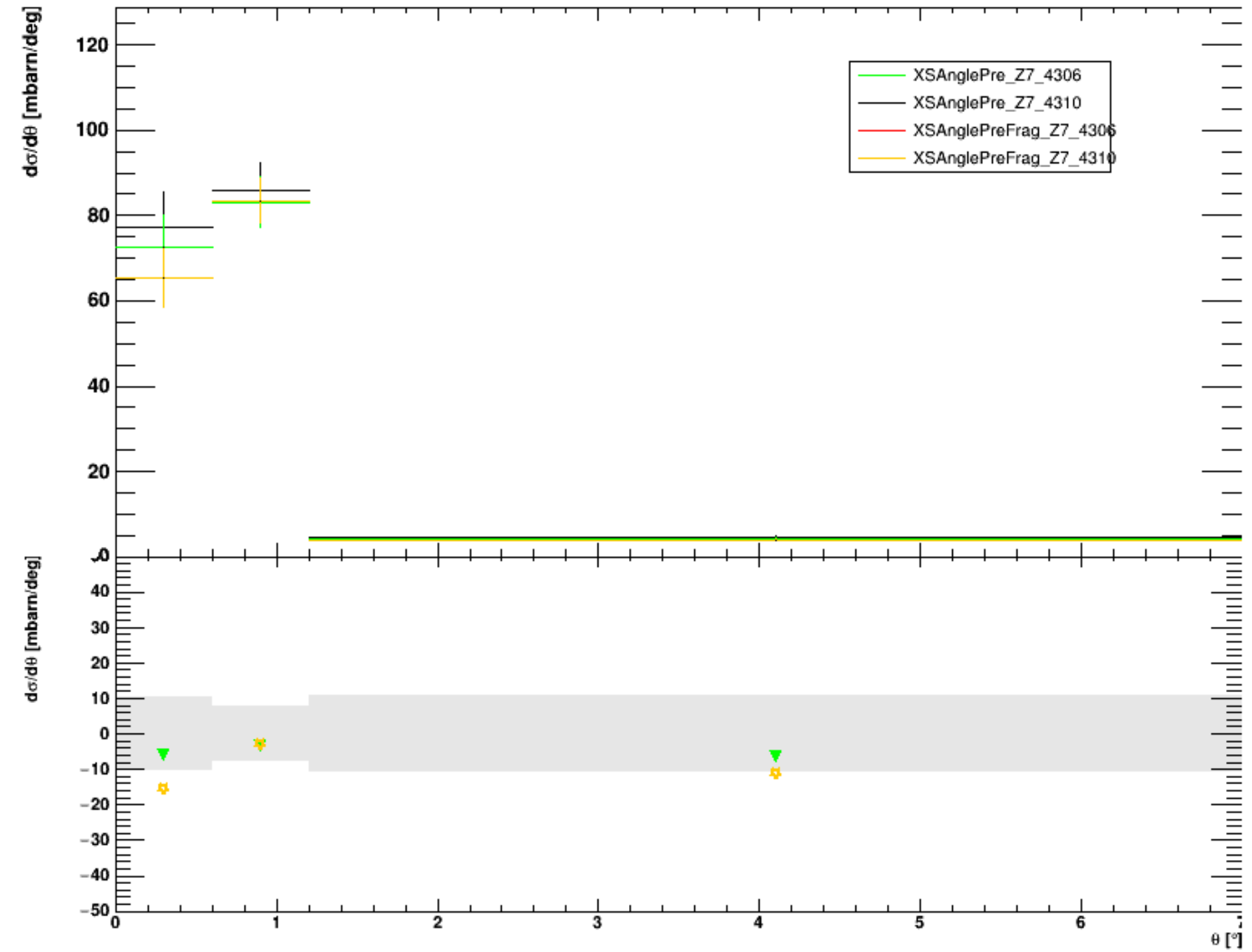
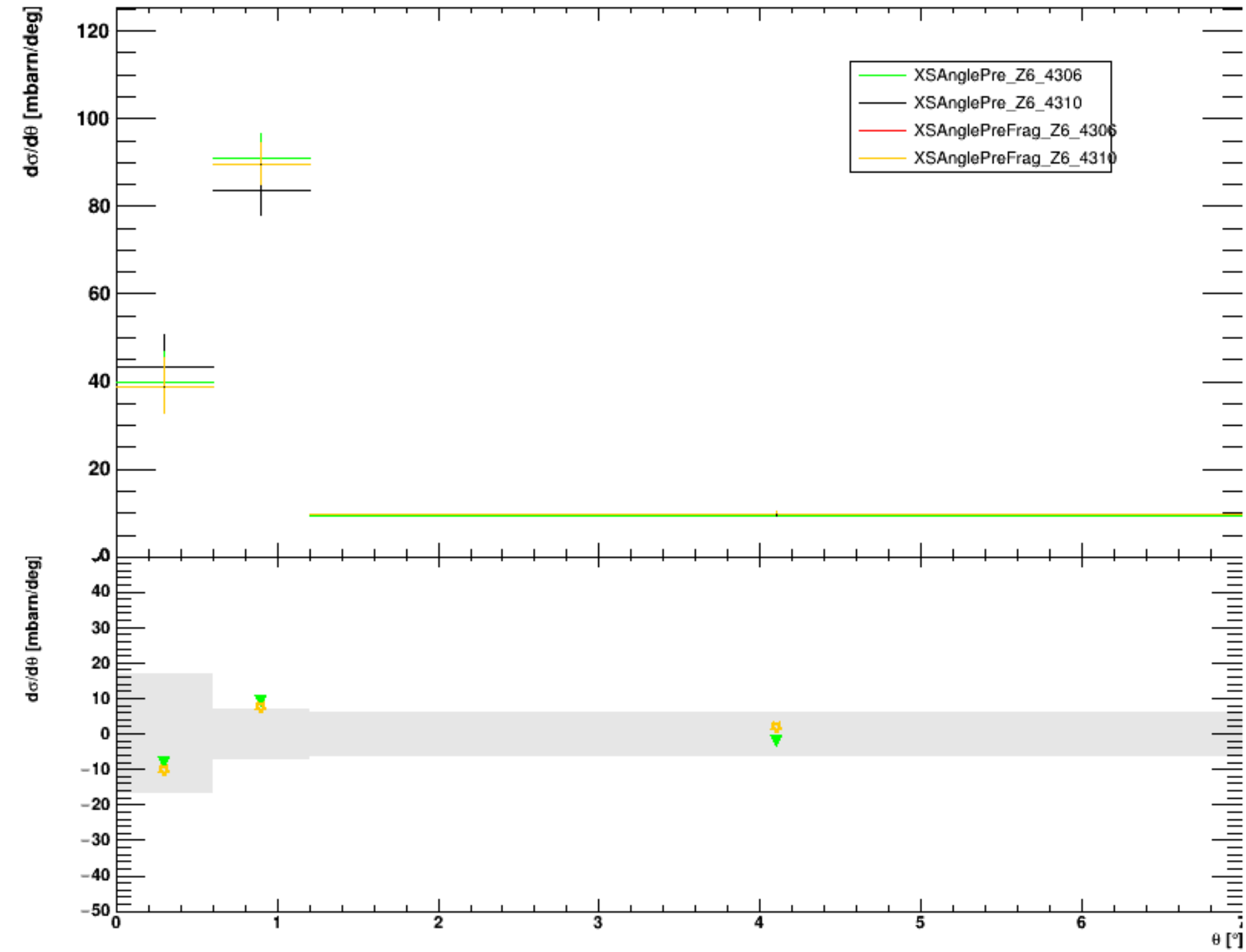
Consistency checks on data



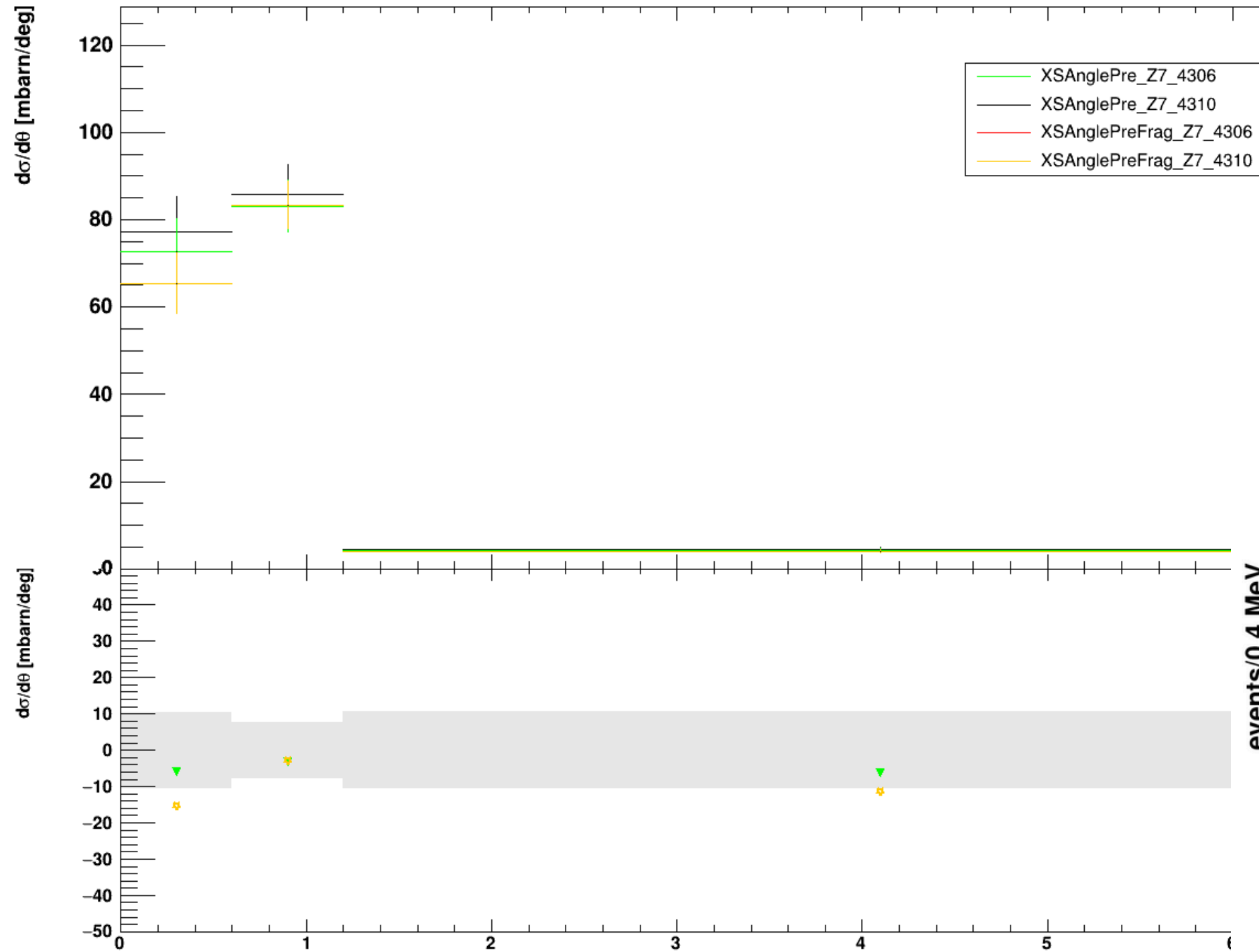
Consistency checks on data



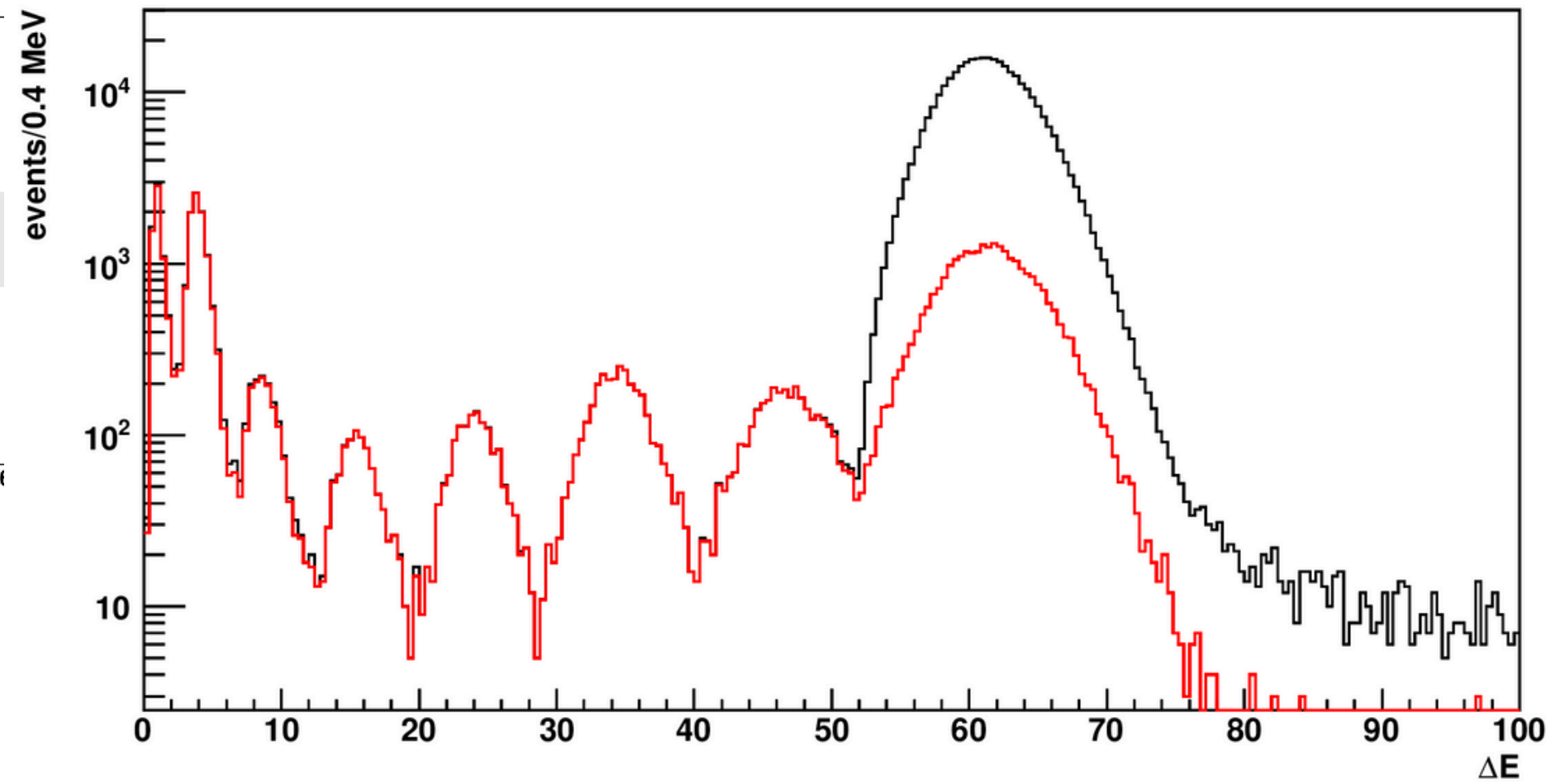
Consistency checks on data



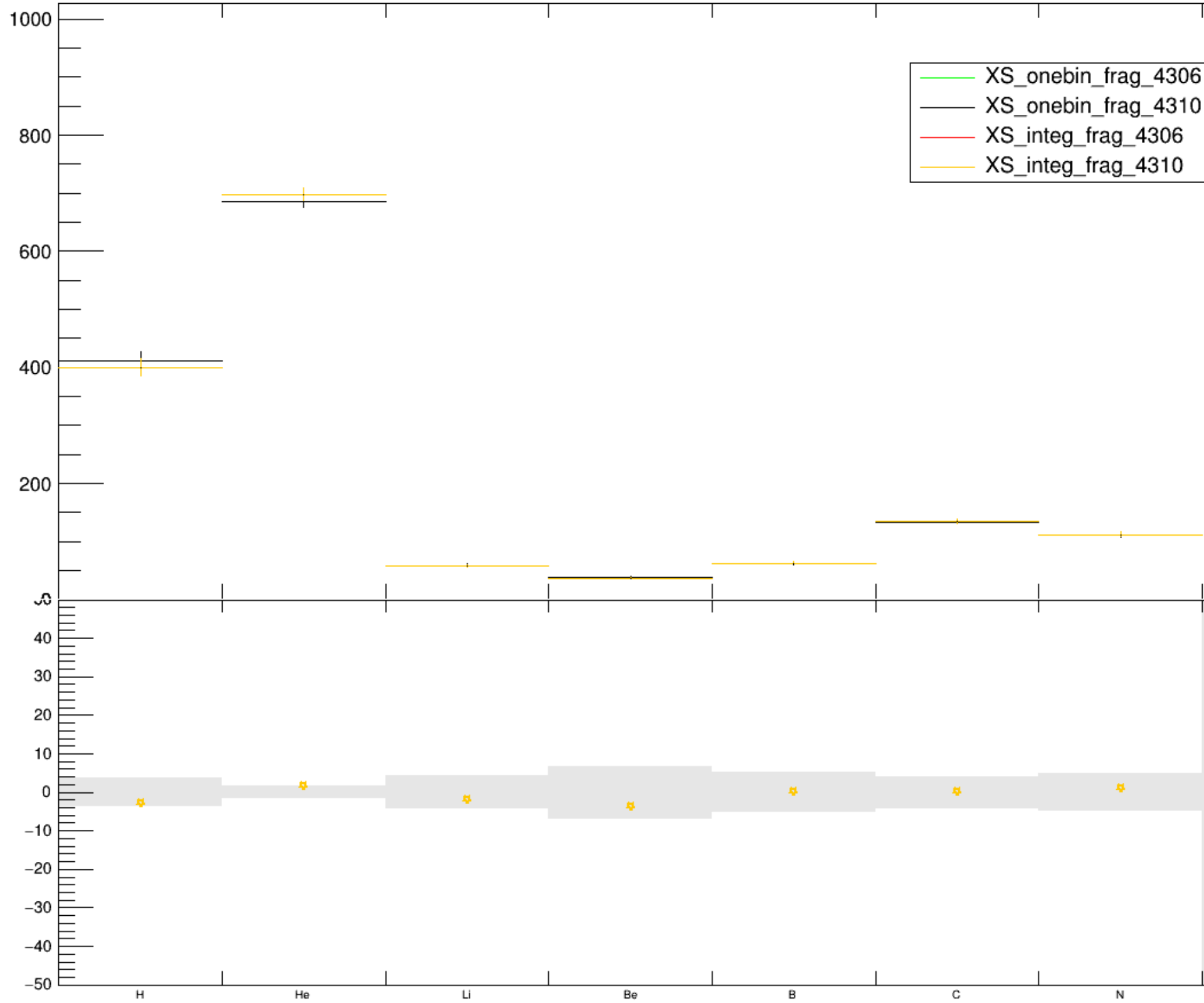
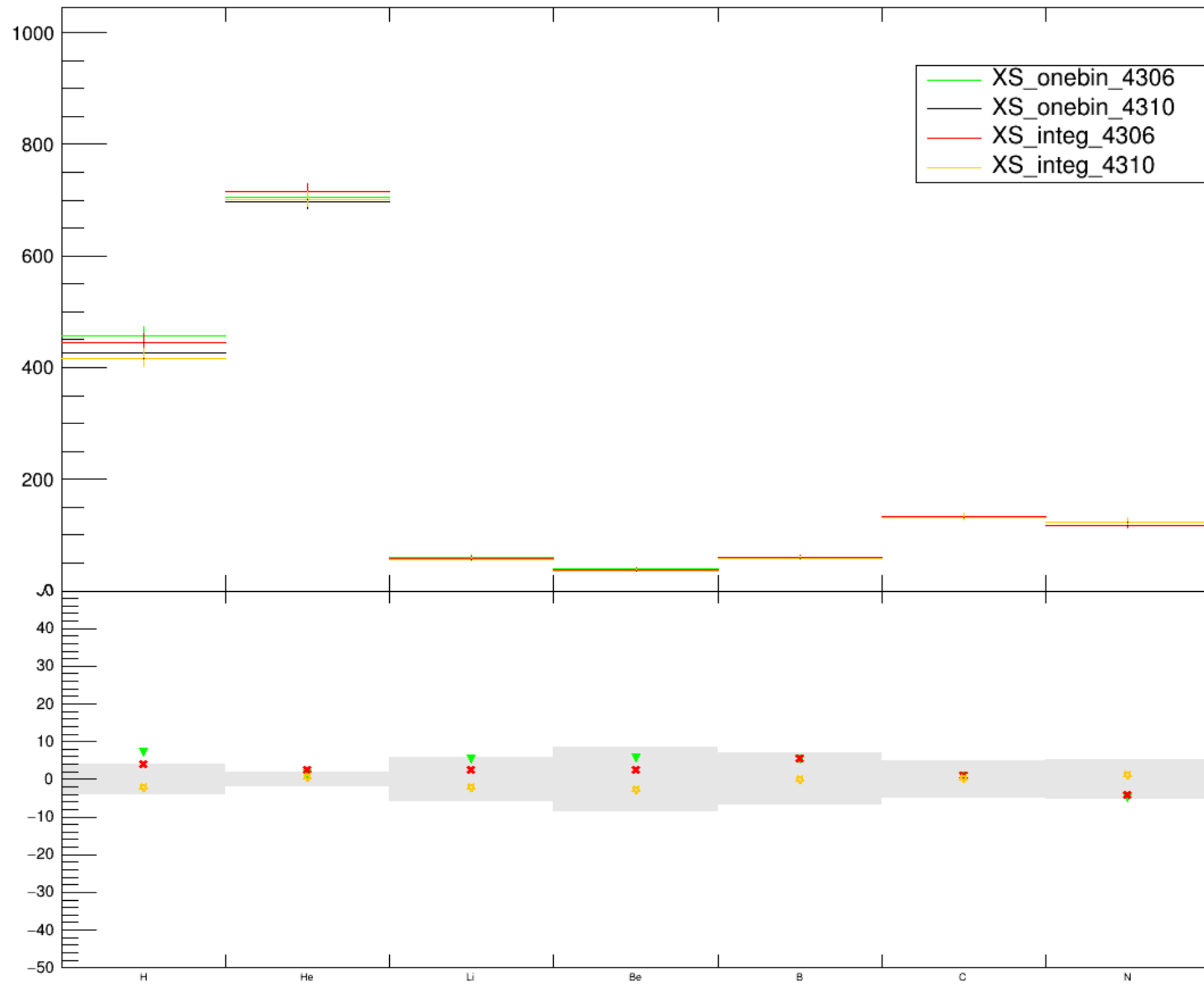
Consistency checks on data



Fragment rescaling maybe useful for N if outside our uncertainties (it seems not)



Consistency checks on data

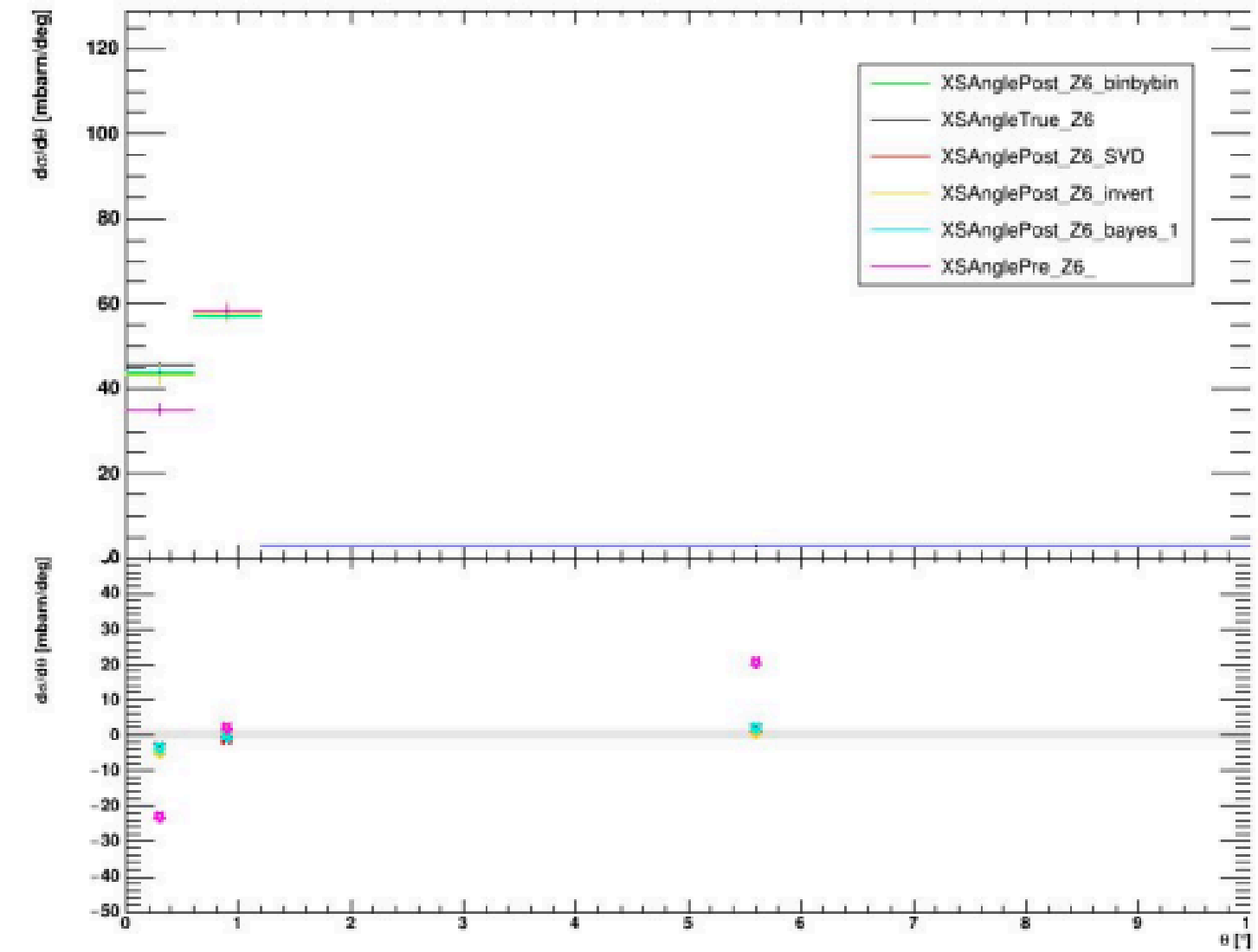
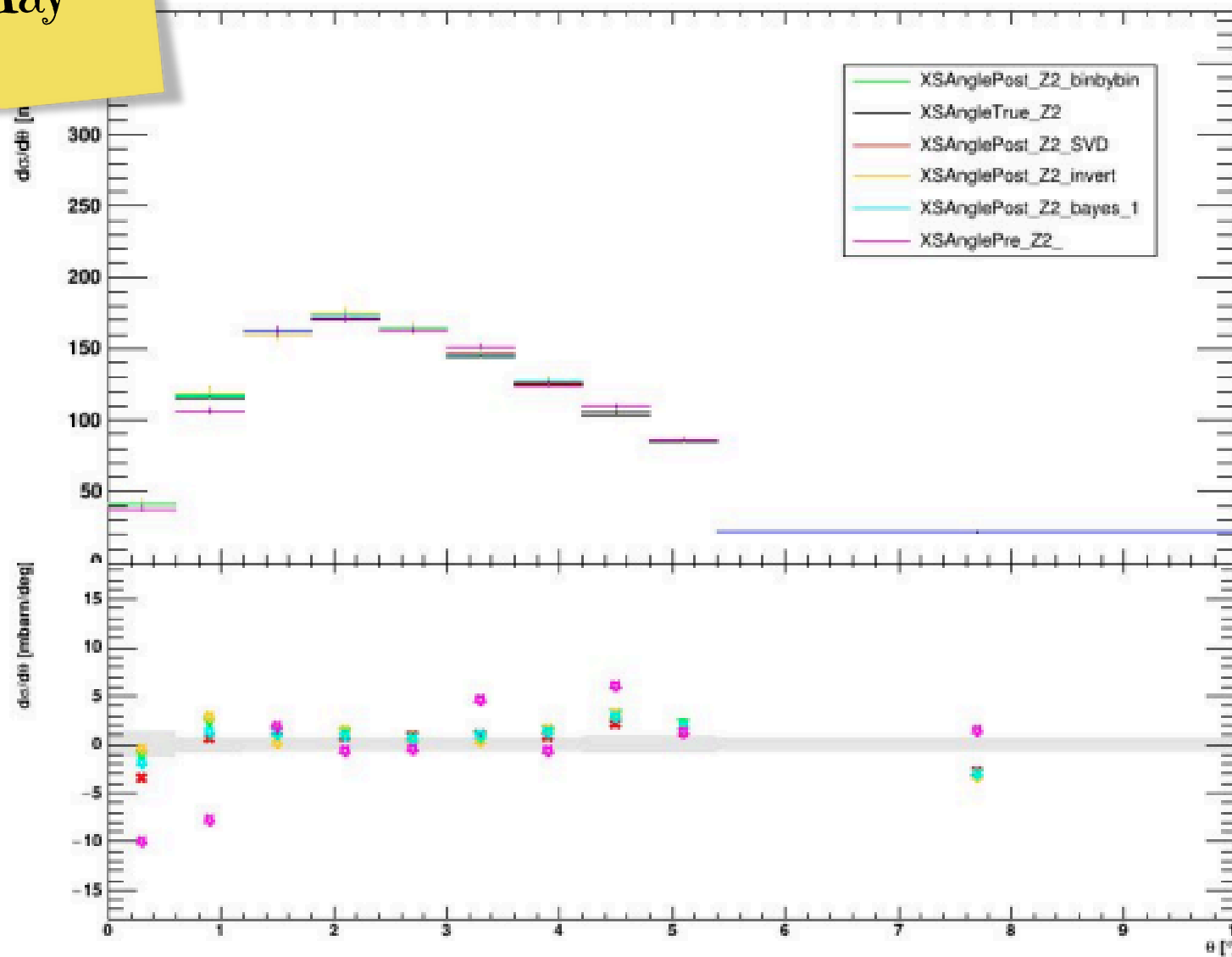


Unfolding Method Comparison

Alberto
21 May

Z=2

Z=6



Different methods behave the same way



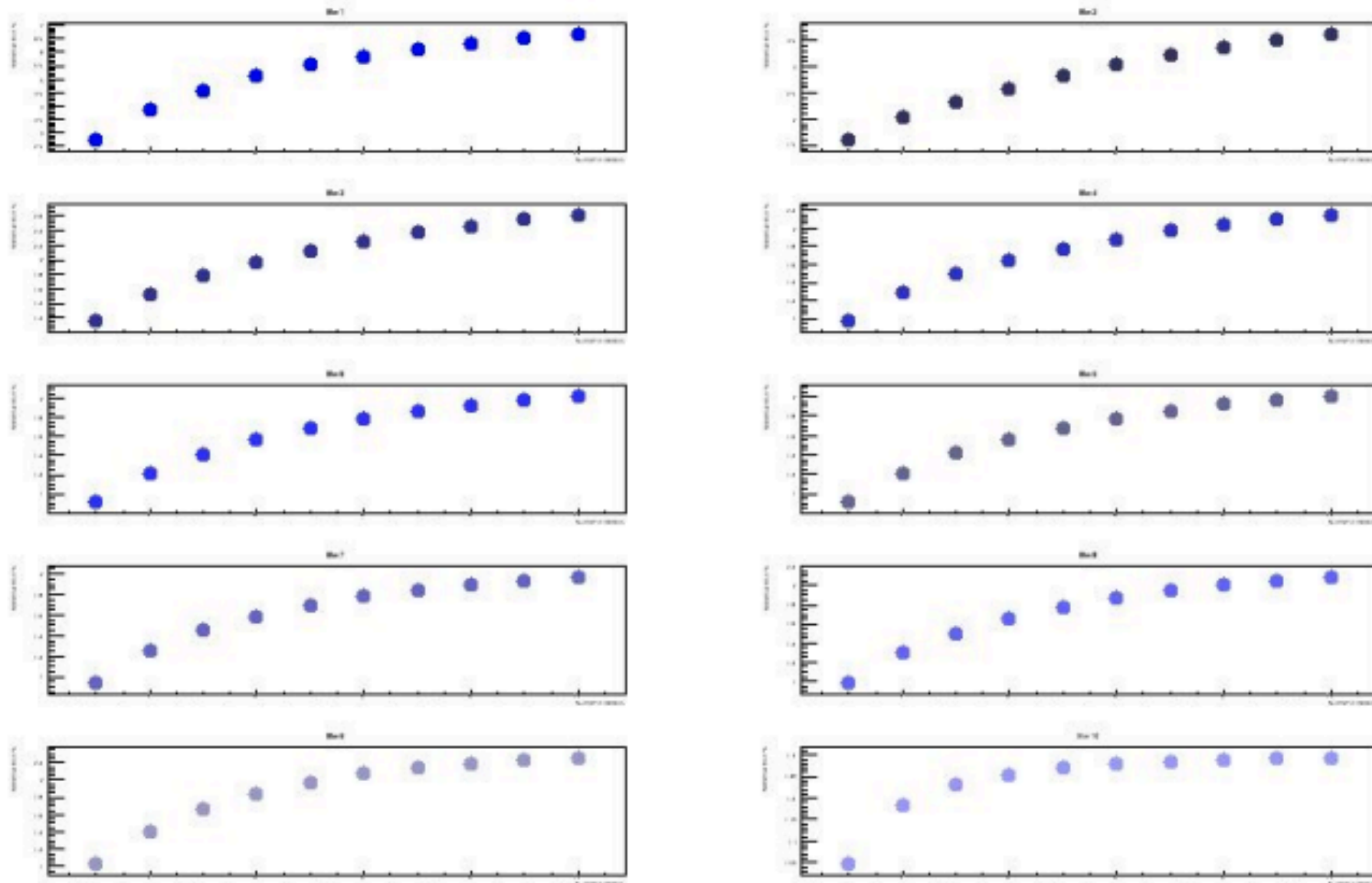
Bayes Iterative method chosen

Alberto
21 May

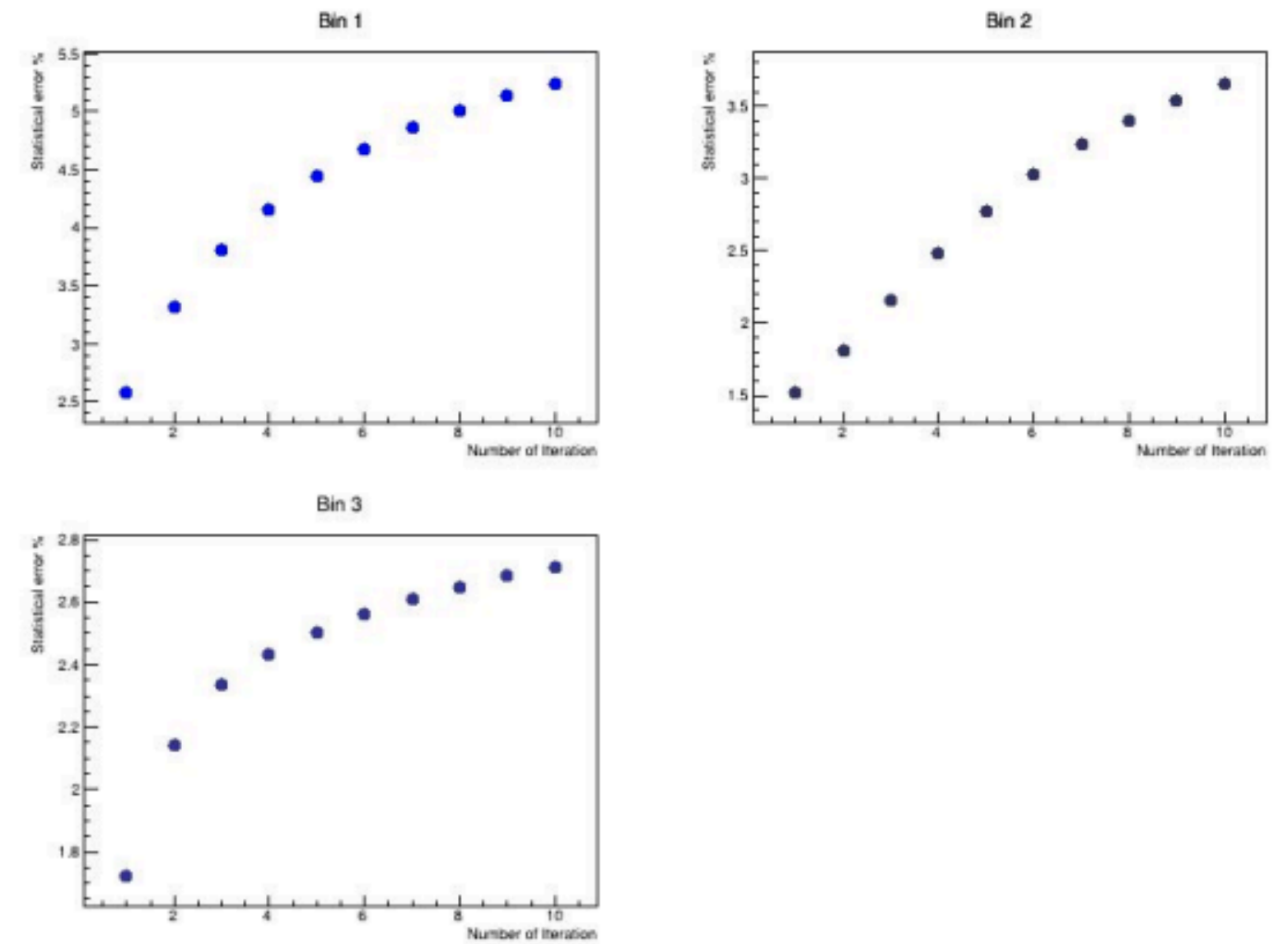
Studies on the number of iterations: Statistical uncertainty

The degree of convergence of the iterative procedure is checked by comparing each iteration to the previous one. In particular, the parameters taken into account are the bin-wise statistical error (it increases with the number of iterations, reaching a sort of plateau)

Z=2



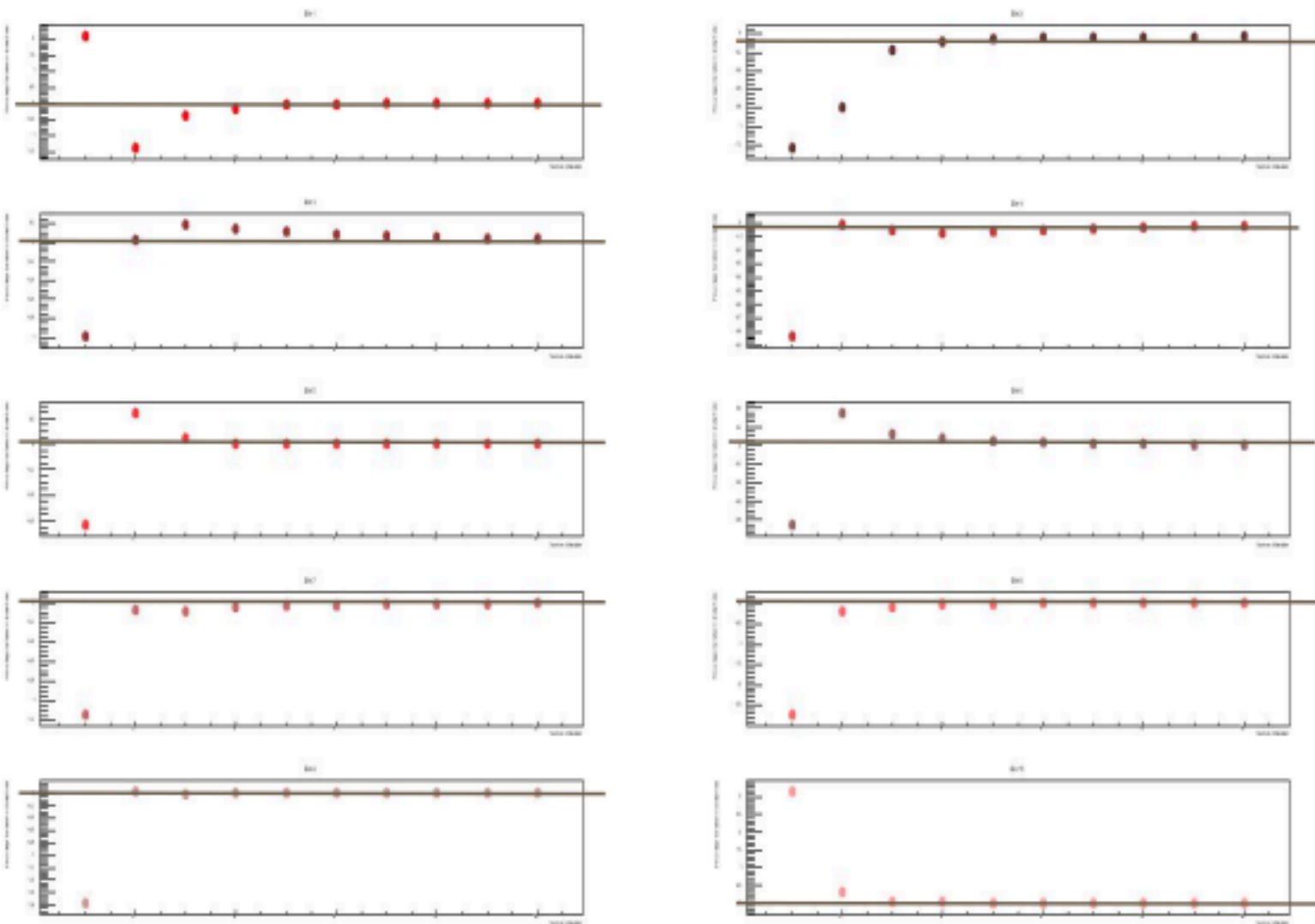
Z=6



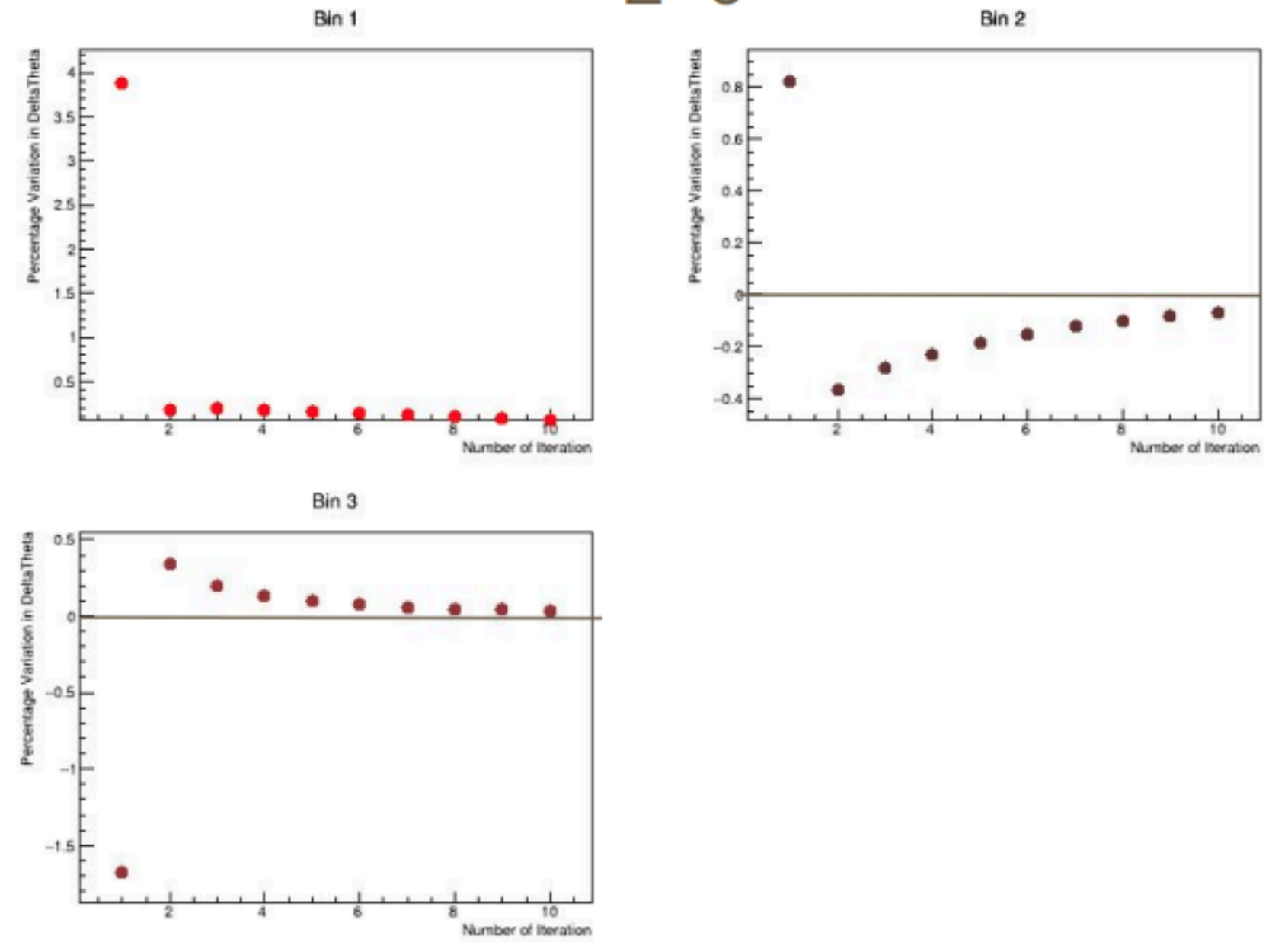
Studies on the number of iterations: bin residuals

bin-wise residuals (the bin-by-bin difference between the unfolded distribution at the i -th iteration and the unfolded distribution at the $i - 1$ -th iteration, which must converge to 0).

Z=2



Z=6



Alberto
21 May

Studies on the number of iterations: Average correlation

An optimal choice of the regularization parameter is the one that minimizes the average correlation factor:

$$\rho_{\text{avg}} = \frac{1}{M_x} \sum_{j=1}^{M_x} \rho_j.$$

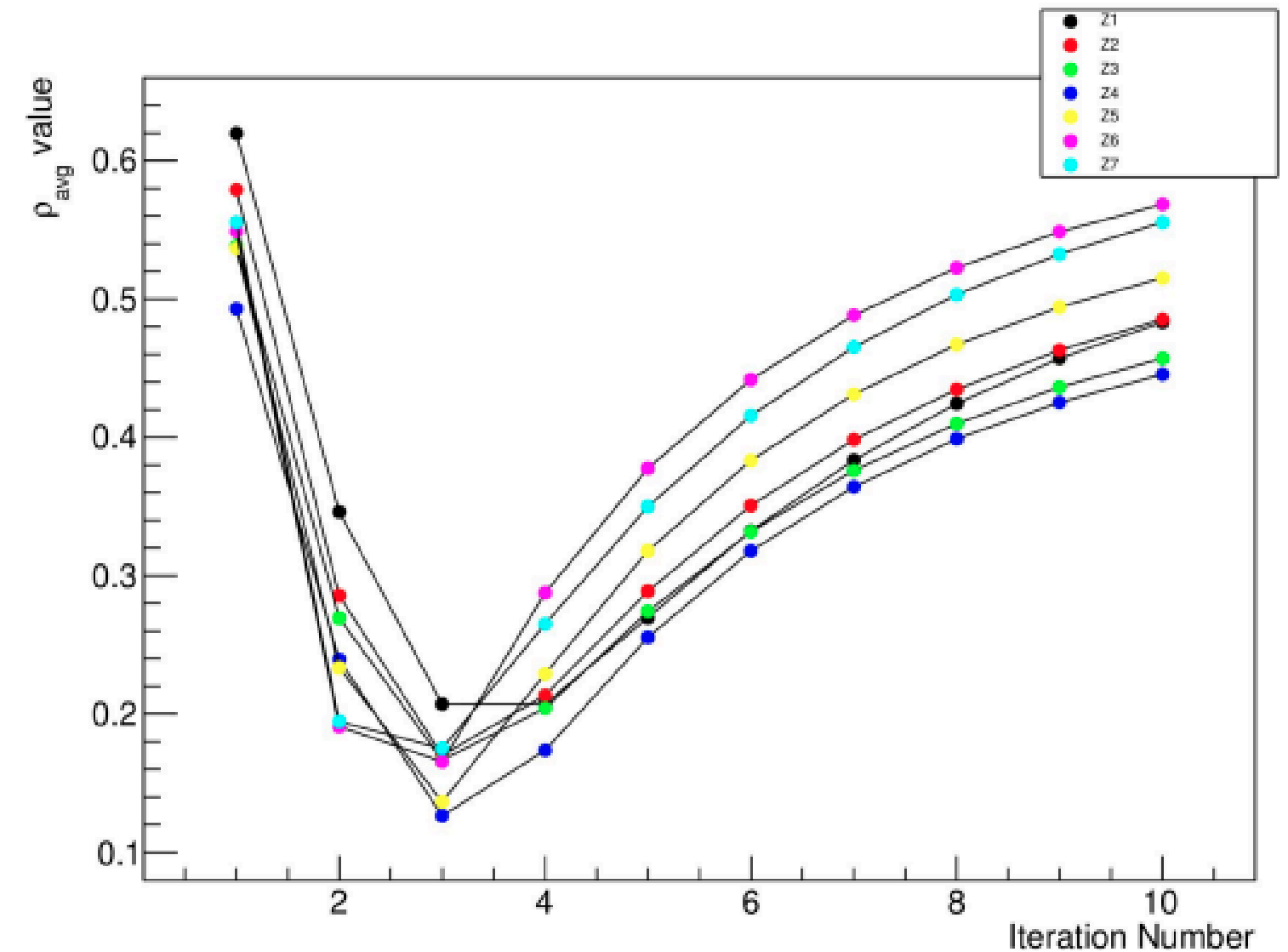
where ρ_j global correlation coefficient of bin j is defined as

$$\rho_j = \sqrt{1 - \left((V_{xx})_{jj} (V_{xx}^{-1})_{jj} \right)^{-1}}.$$

M_x : ndof; V_{xx} : Cov. matrix

$$\rho_{\text{avg}} = \frac{1}{N_{\text{bins}}} \sum_{i=1}^{N_{\text{bins}}} \rho_i \quad \rho_i = \sqrt{1 - [C_{ii} (C^{-1})_{ii}]^{-1}}.$$

S. Schmitt, Data Unfolding Methods in High Energy Physics, EPJ Web Conf. 137 (2017) 11008, ed. by Y. Foka, N. Brambilla and V. Kovalenko, arXiv: 1611.01927

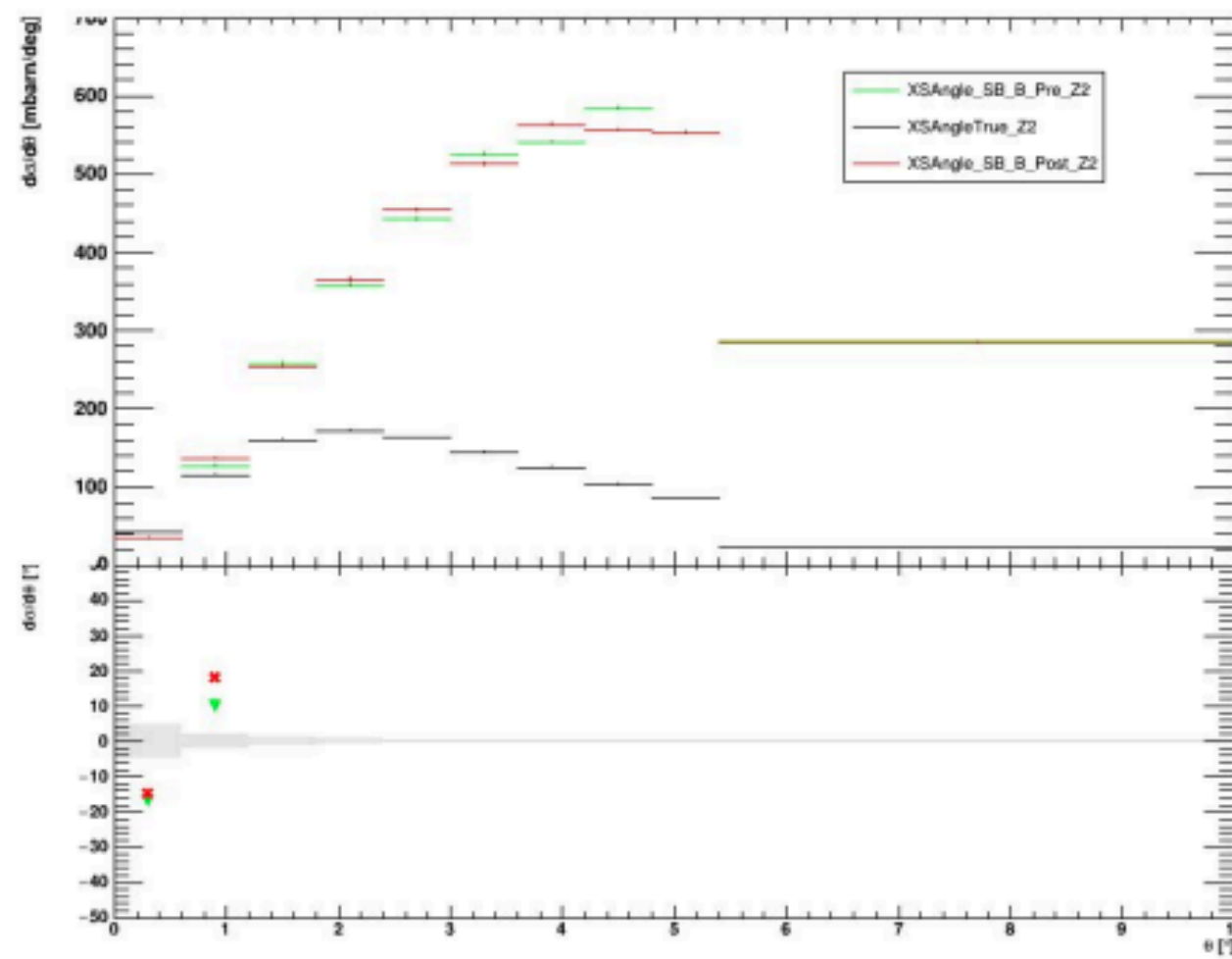


Alberto
21 May

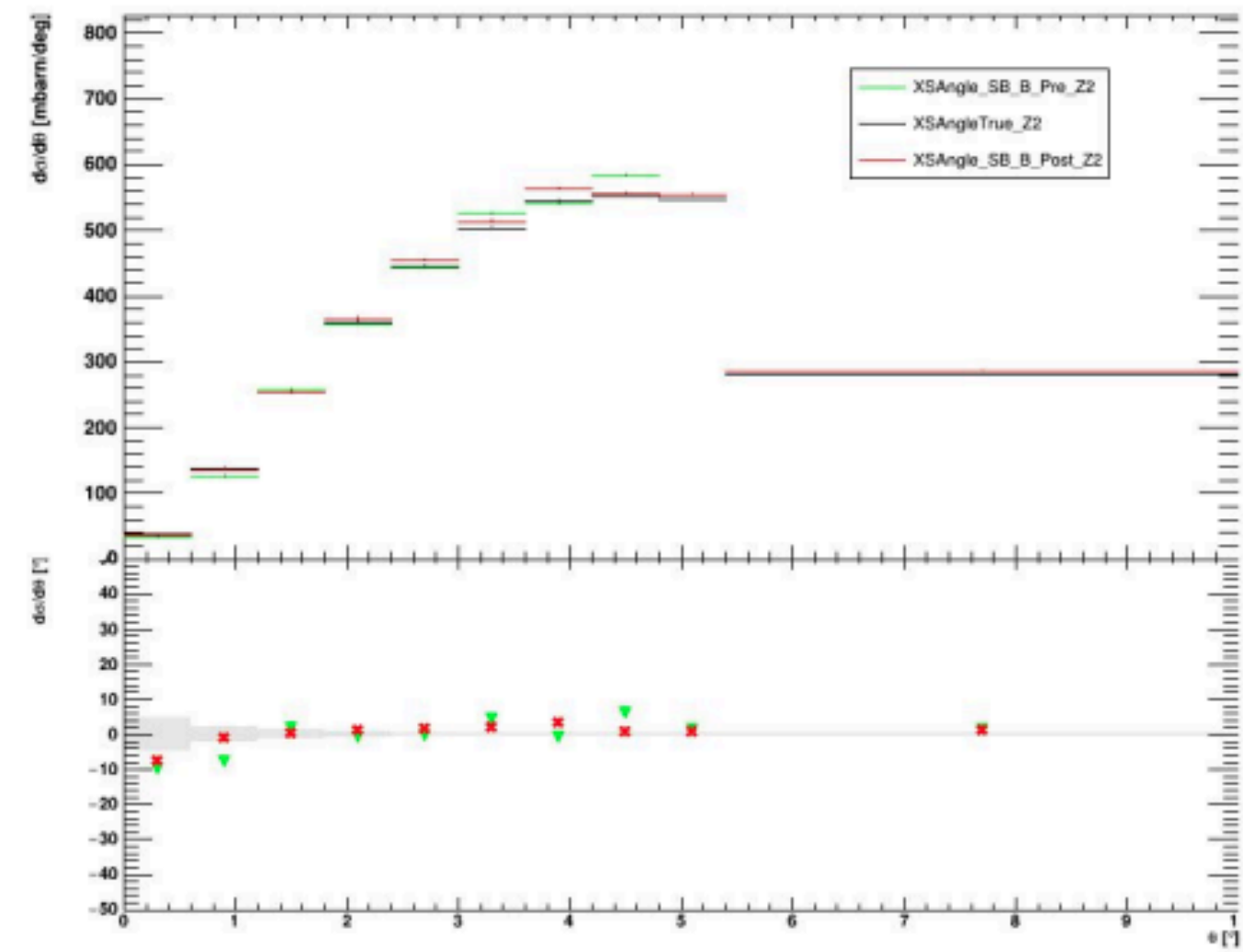
Stress tests

Due to the specific choice of the Monte Carlo sample for the training of the unfolding, it is necessary to check whether this choice could introduce a bias via the unfolding. To do this check the MC reweighting is required, in order to change the shapes of the distributions and get a varied distribution used as pseudo-data.

Only pseudo-data reweighted

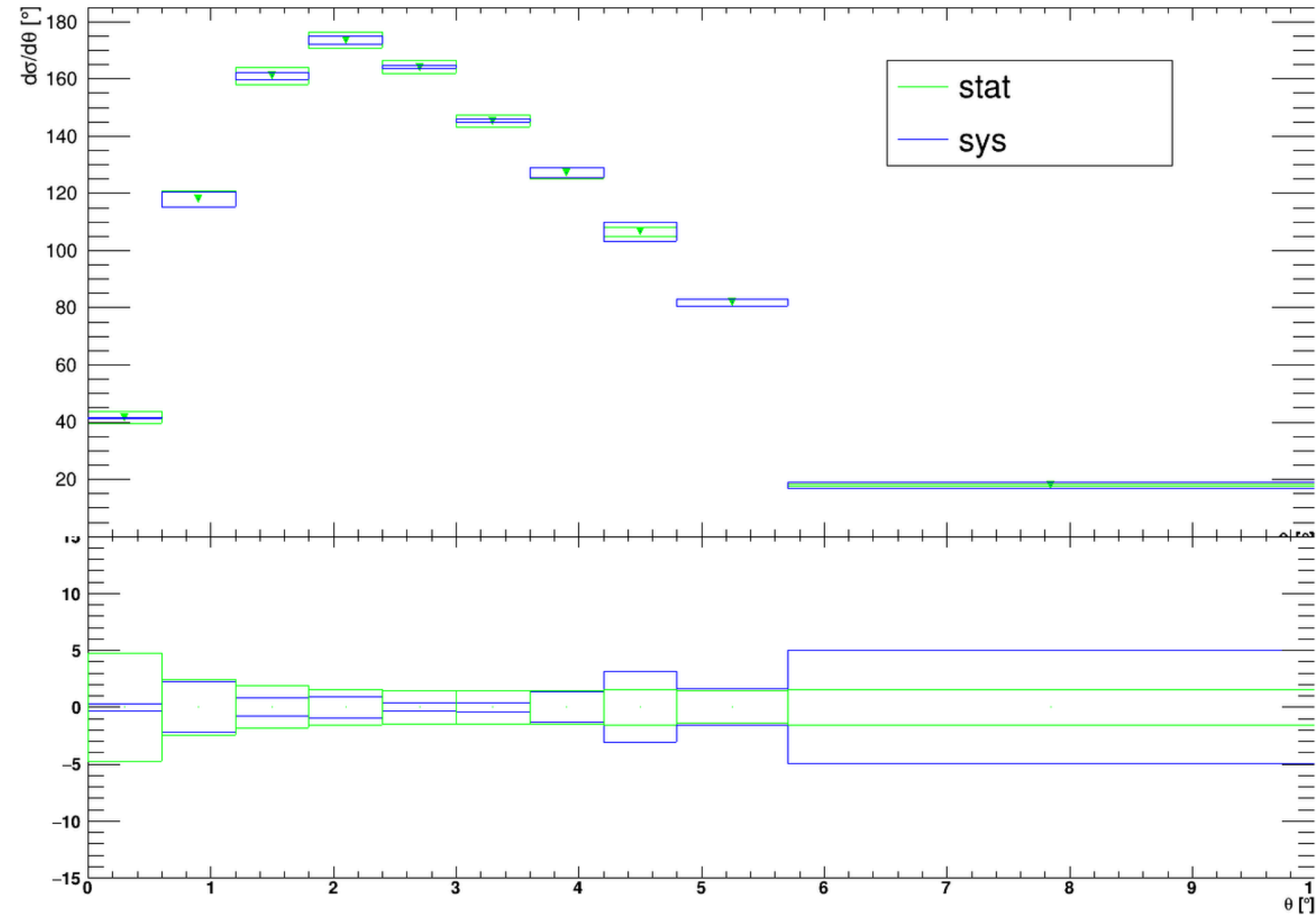


Both pseudo-data and truth reweighted

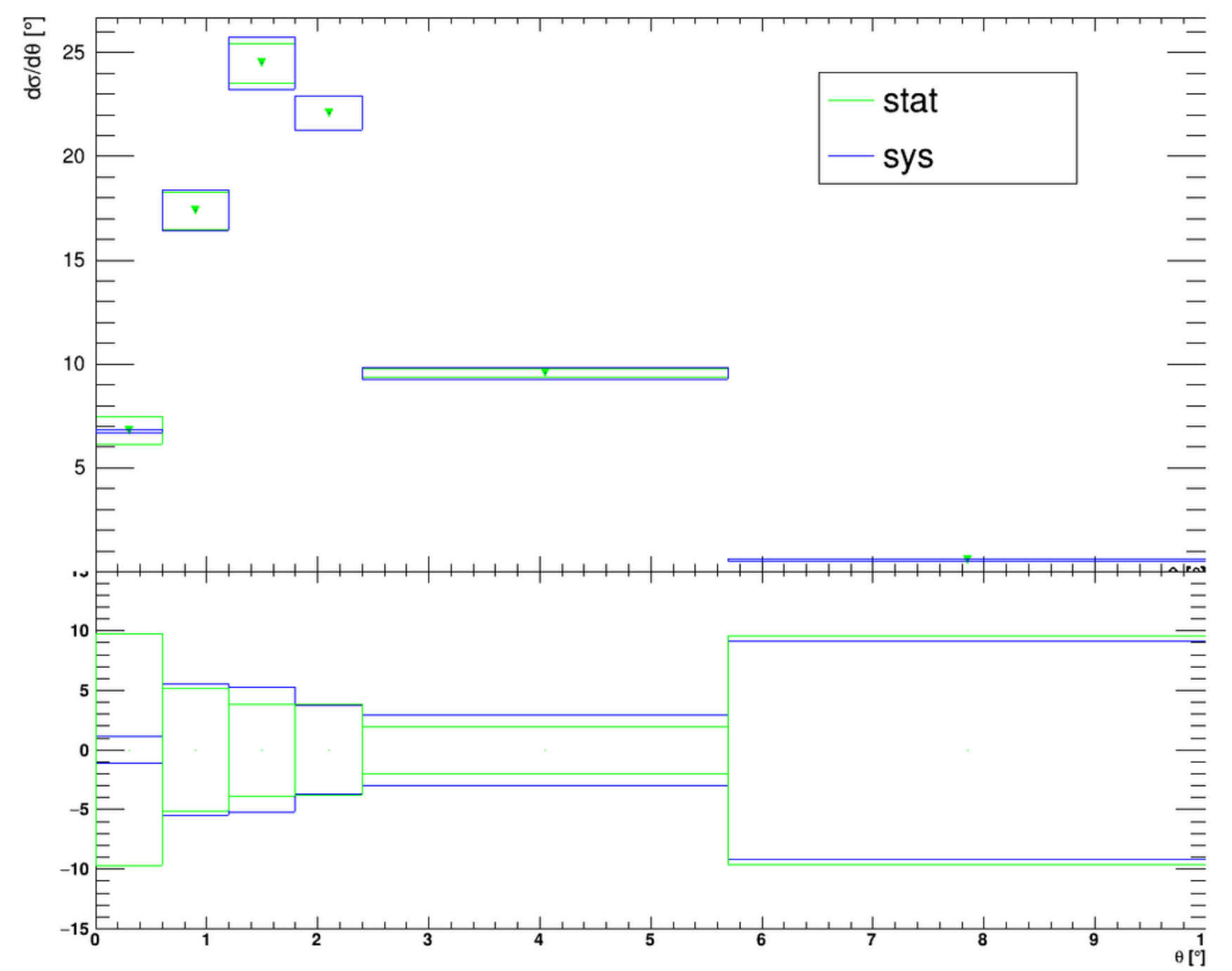


Merge all statistics

Z=2



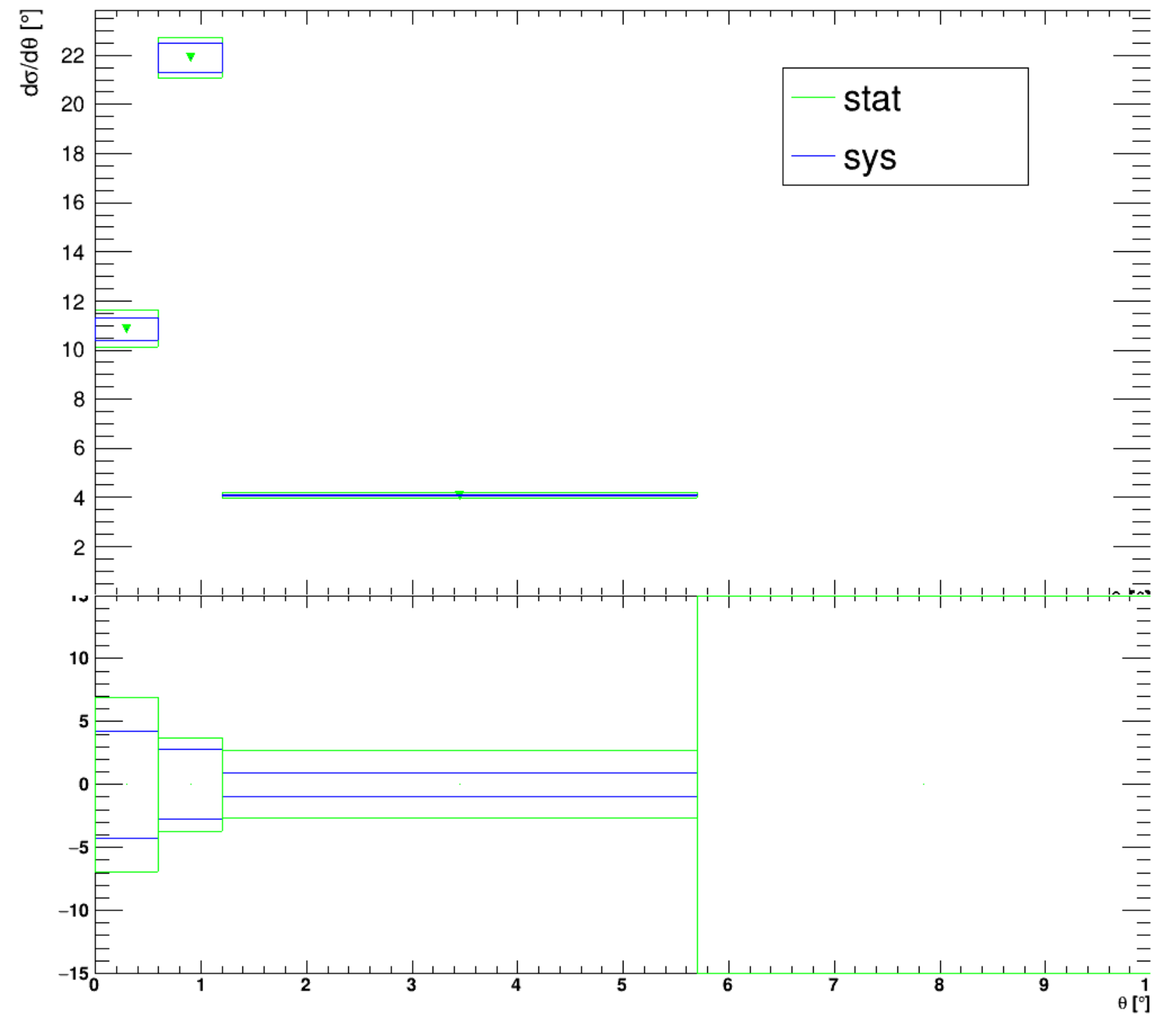
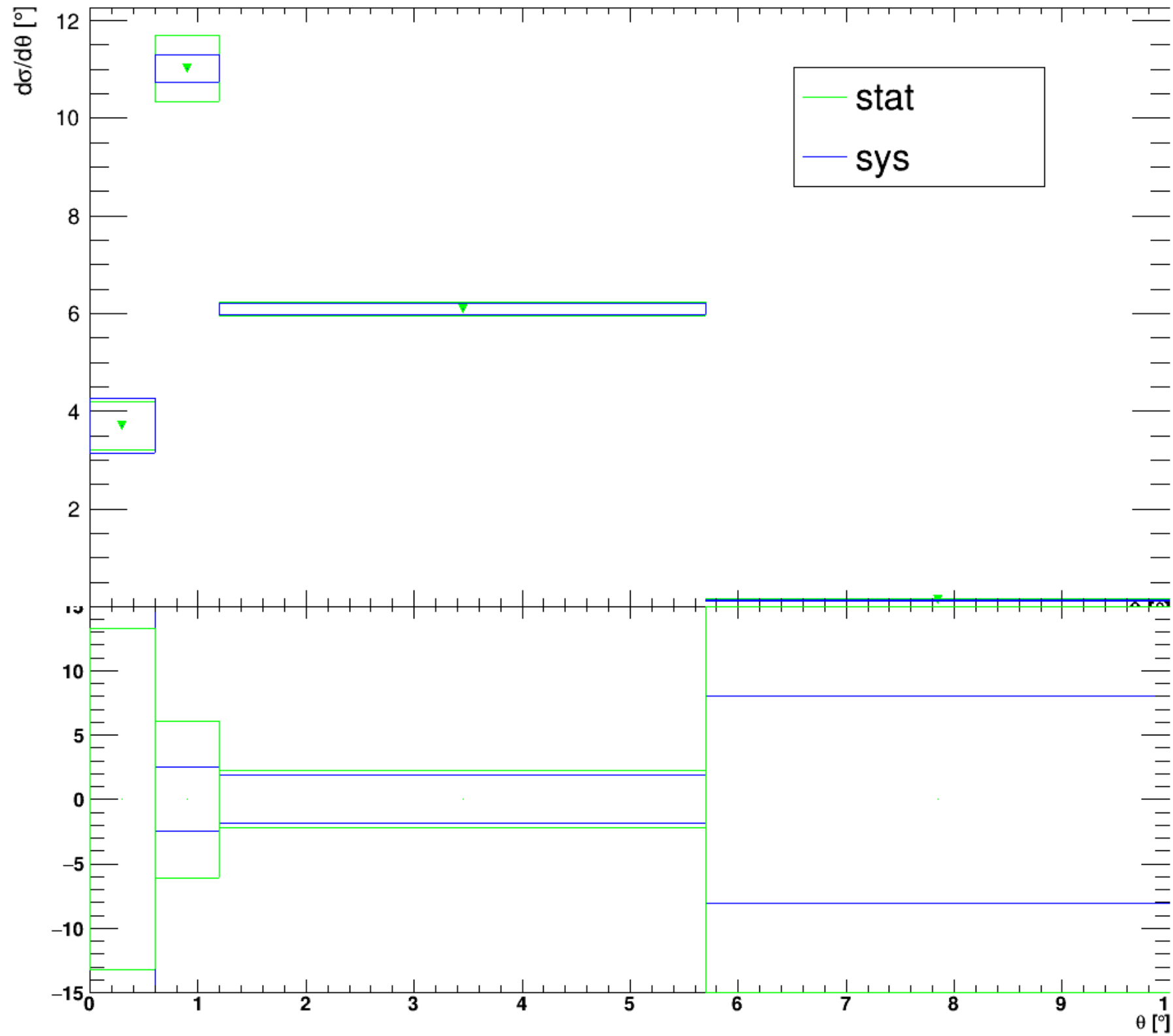
Z=3



Merge all statistics

Z=4

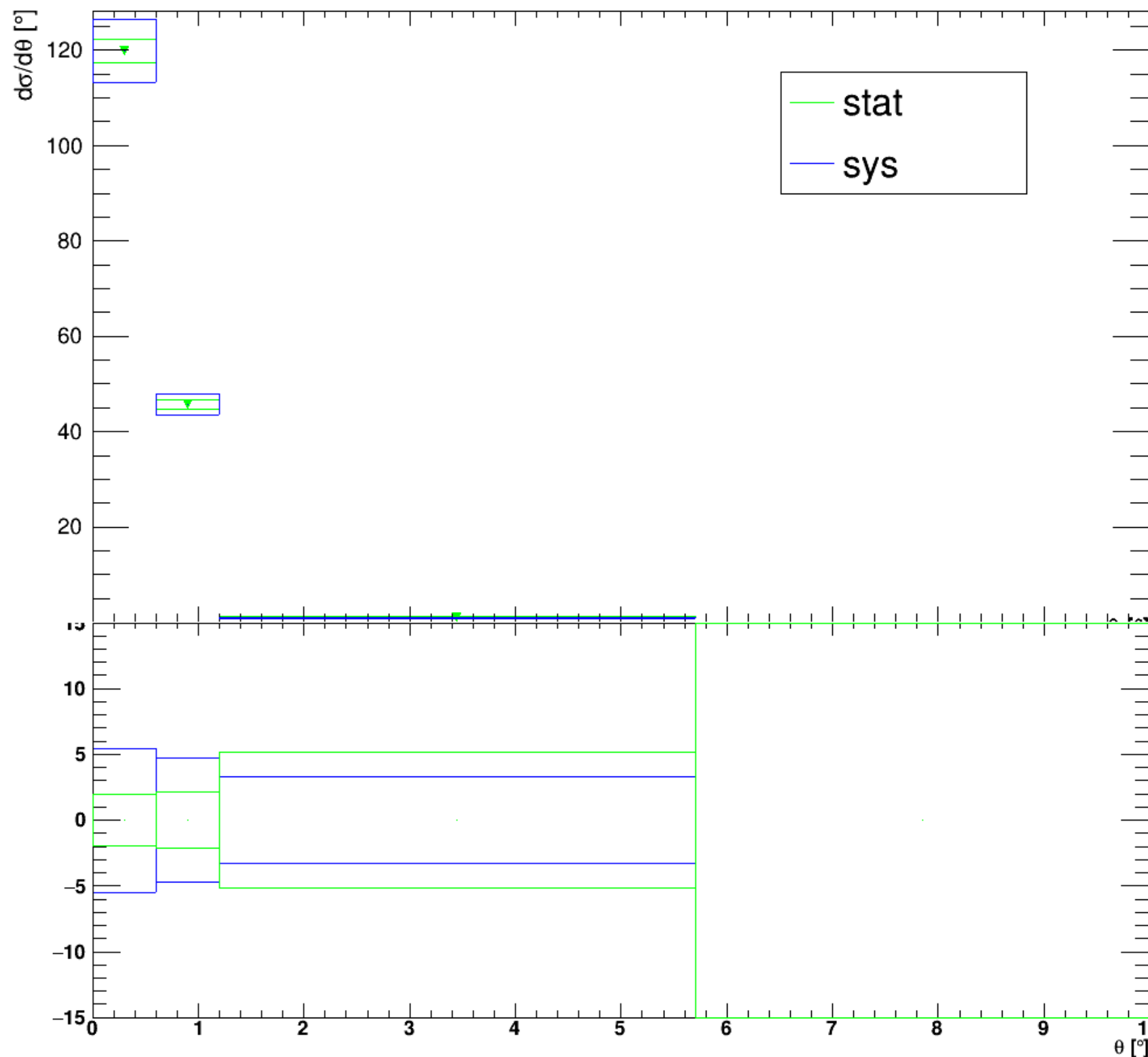
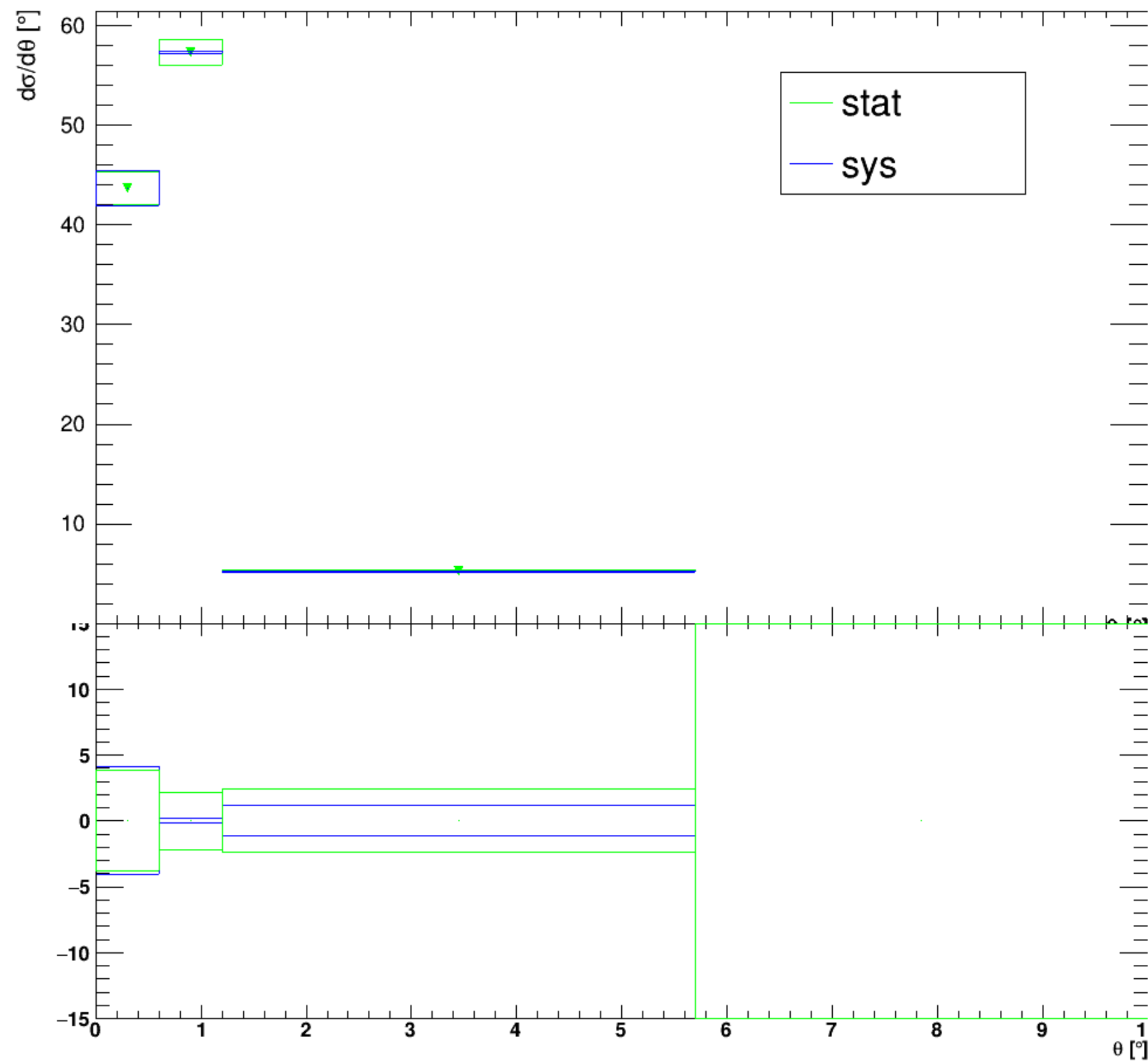
Z=5



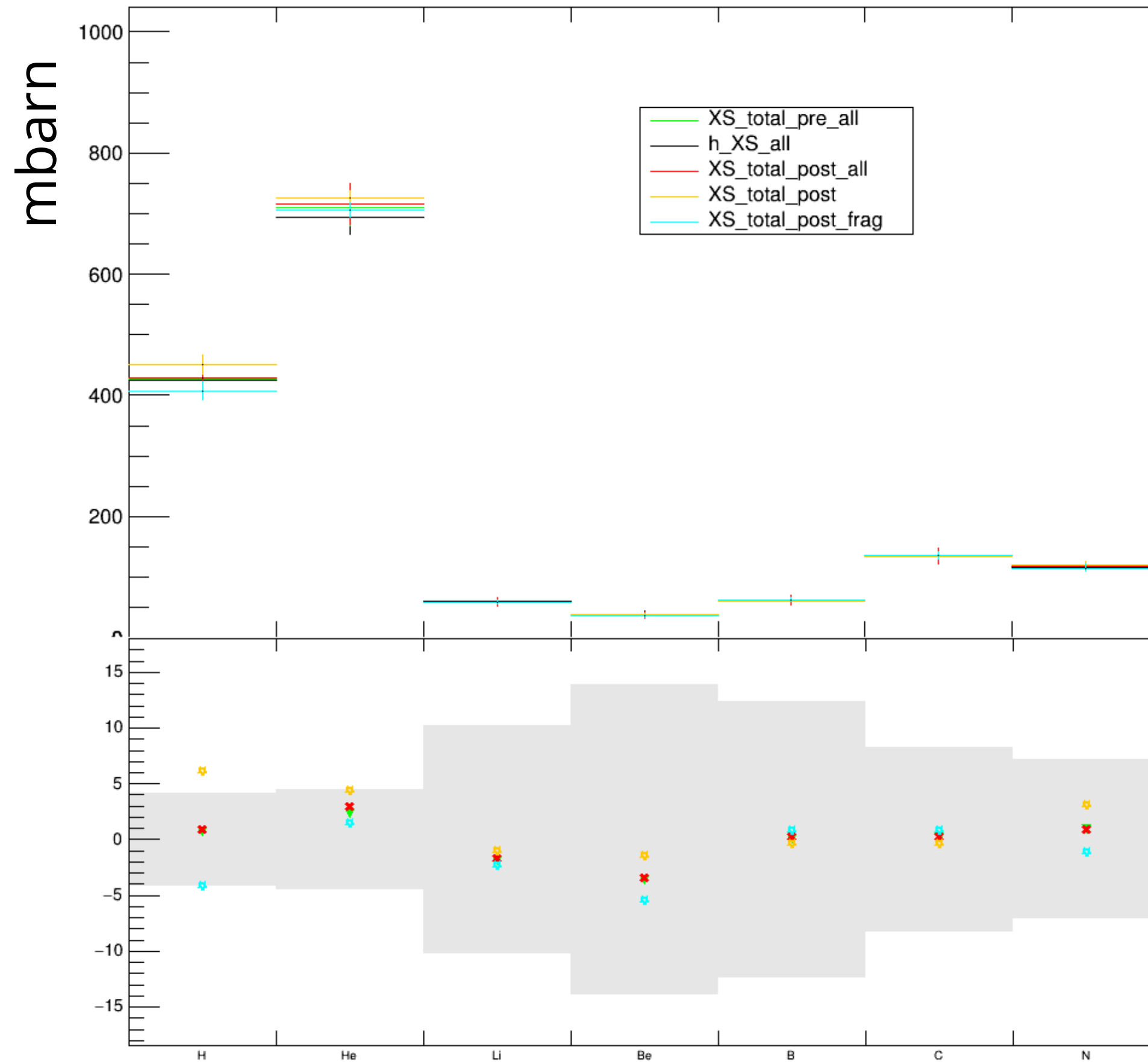
Merge all statistics

Z=6

Z=7



Merge all statistics



Next steps

Data seem to agree among runs

Merged all the statistics with target

Added systematics of the subtraction method

Bayes selected as unfolding method, others in systematics

Geometric efficiency to add for $Z=2$

We can start writing!

Thanks for listening!