

Superiteration for unfolding

(Following J/Psi Analysis)

Flowchart:

- Start with a flat prior in fragmentation function (Z)
 - Unfold the 2D jet pT vs Z distribution with flat prior assumption
 - Use the unfolded Z distribution as a prior in next step
 - Terminate at a reasonable chi-square value
- x repeat n times -----> Calling this “**SUPERITERATION**”

Example:

1. Choose number of iterations for each unfolding → Usual value is 4 (I vary it from 3 to 20)
2. Start with a flat Z distribution and unfold the 2D jet pT vs Z detector level distribution
3. The unfolded Z distribution is used as a prior in the next step.
4. Unfold the original 2D jet pT vs Z detector level distribution with the new response matrix

Steps 2 and 3 are called a **SUPERITERATION**

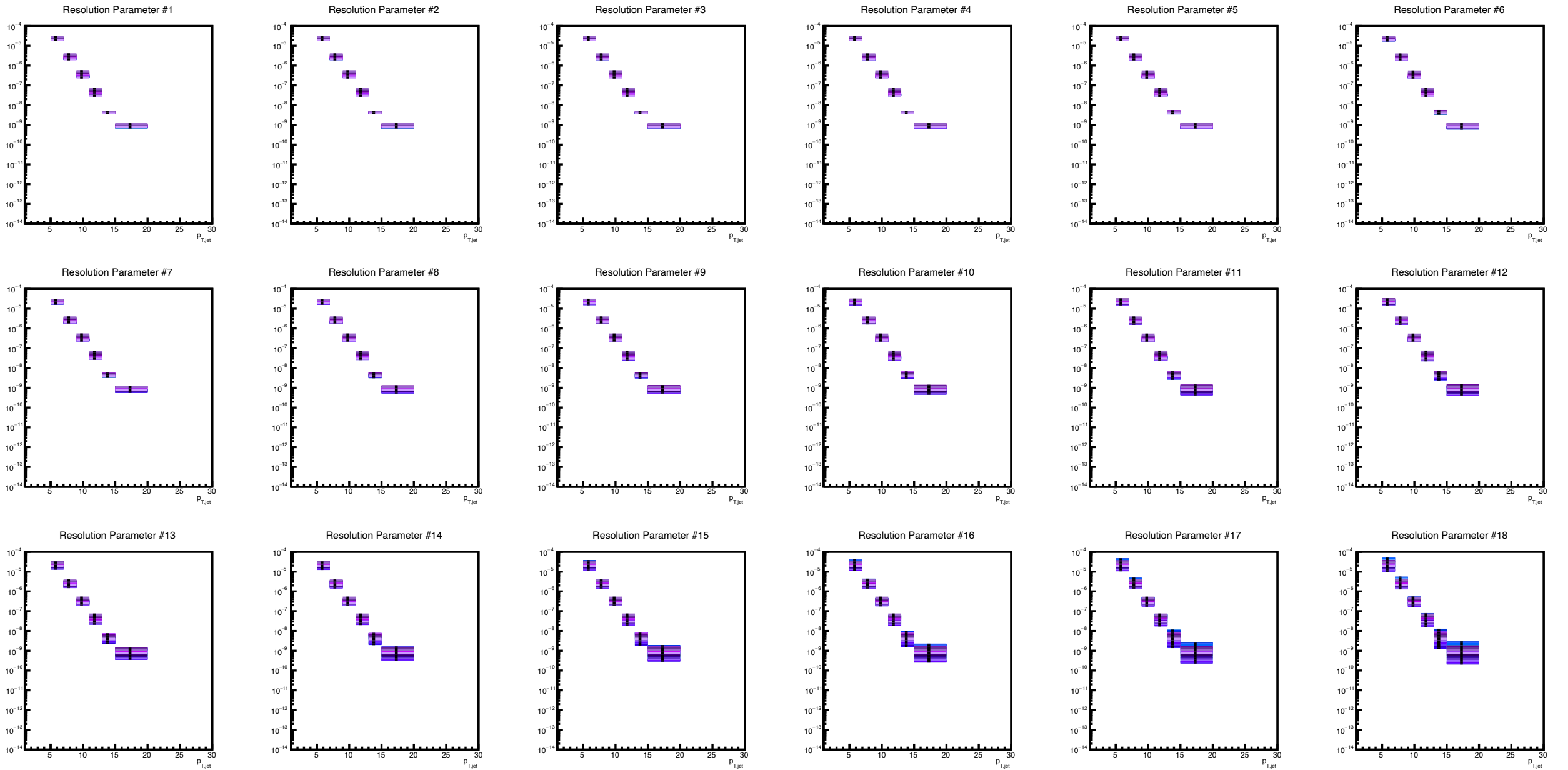
Chi-Square (summed over all bins):

$$\chi^2 = \sum \frac{(\text{Observed} - \text{Expected})^2}{\text{Expected}}$$

For self closure, this is straight-forward, because we know the expected value (PYTHIA Truth)

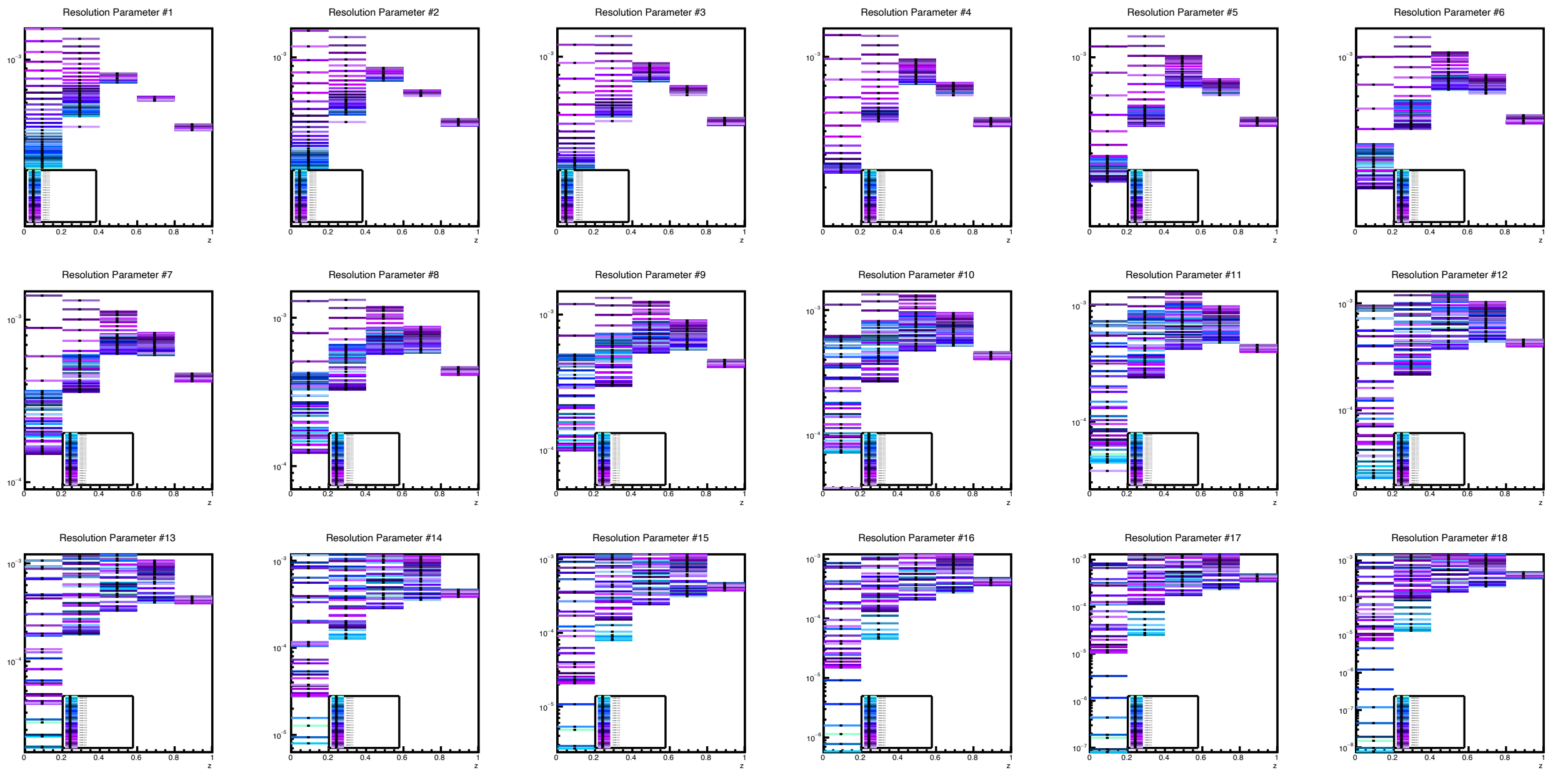
For data, I use the PYTHIA truth as the expected value.

The jet p_T spectra is recovered in most cases, which is clear in the ratio plots in the next slide



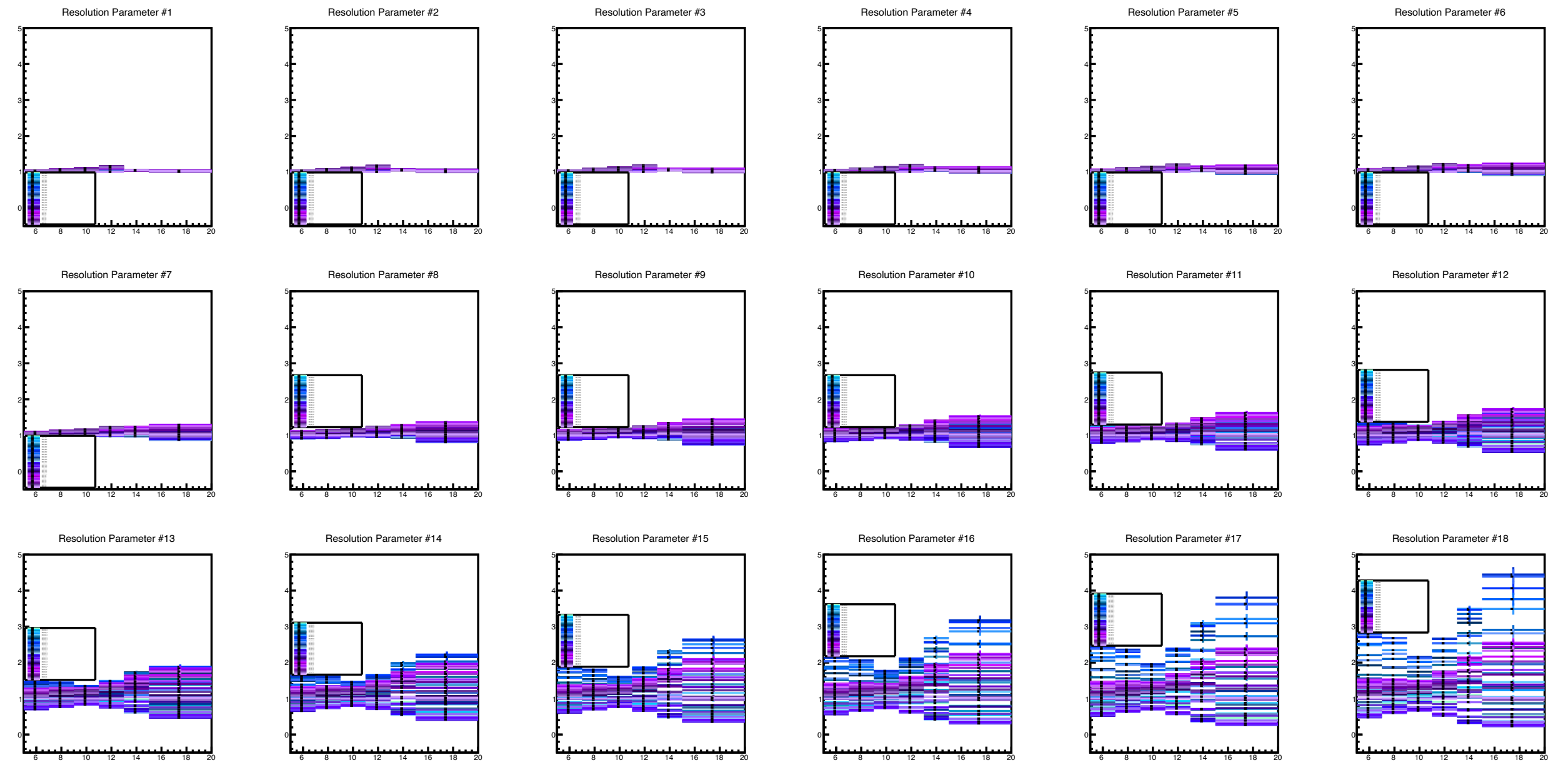
Number of superiterations on the x-axis

The jet Z spectra drastically changes with superiterations

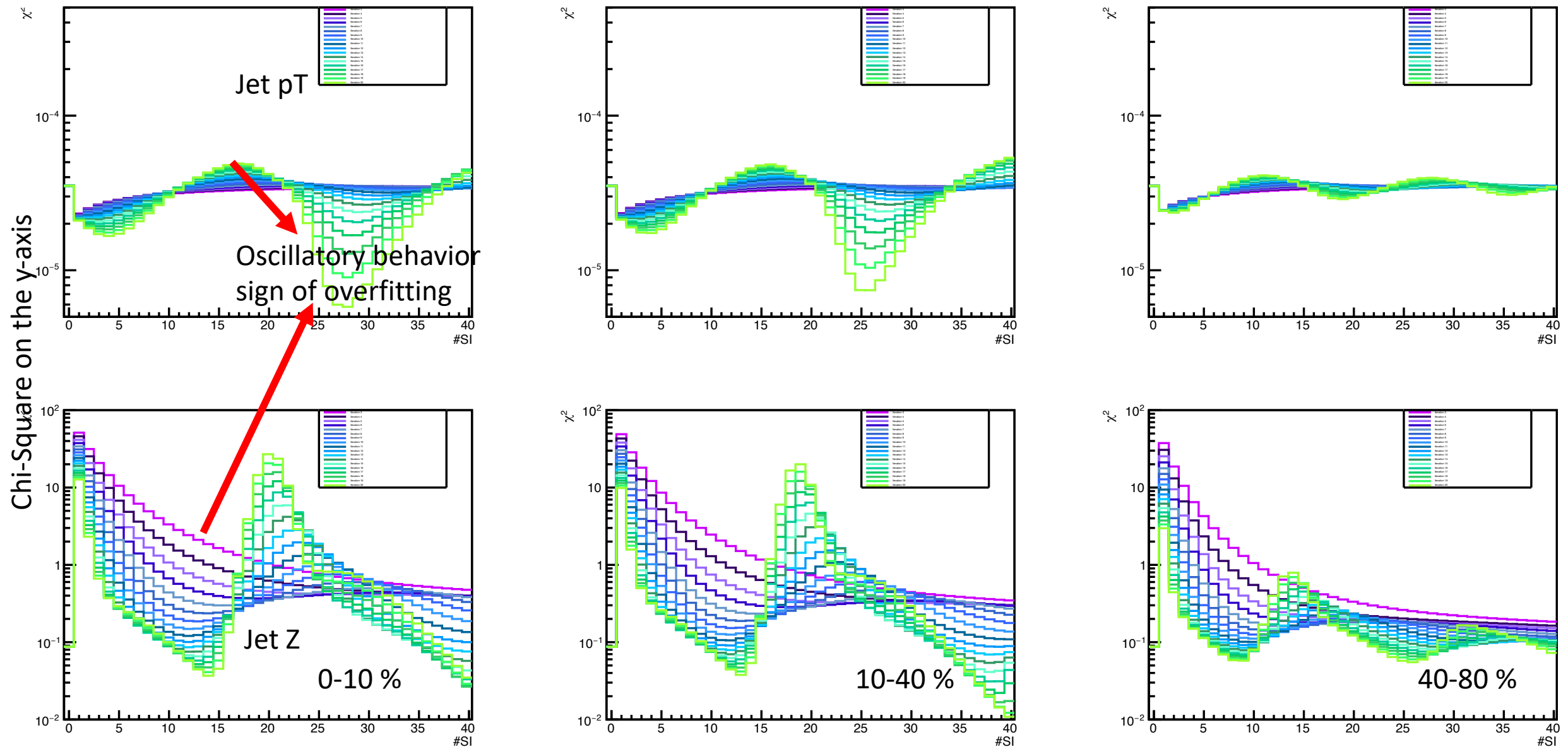


Number of superiterations on the x-axis

The method can get back the PYTHIA jet p_T spectra for reasonable values of the resolution parameter and superiterations



Number of superiterations on the x-axis



Steps for unfolding data:

For Each Centrality Bin

1. Make 4D response matrix (pT vs Z) from 0-3 and 3-inf pthatbins, and combine them with cross-section weights
2. Reweight fragmentation function distribution
 - **Hist1**: Normalise the uncorrected z-distribution from data to have integral 1.
 - **Hist2**: Normalise the detector side PYTHIA z-distribution to have integral 1.
 - Weight \rightarrow Hist2/Hist1
3. Repeat Steps 1 and 2 \rightarrow **Superiteration** (SI)
4. After unfolding, scale pT distribution by T_{AA}

Chi-Square (summed over all bins):

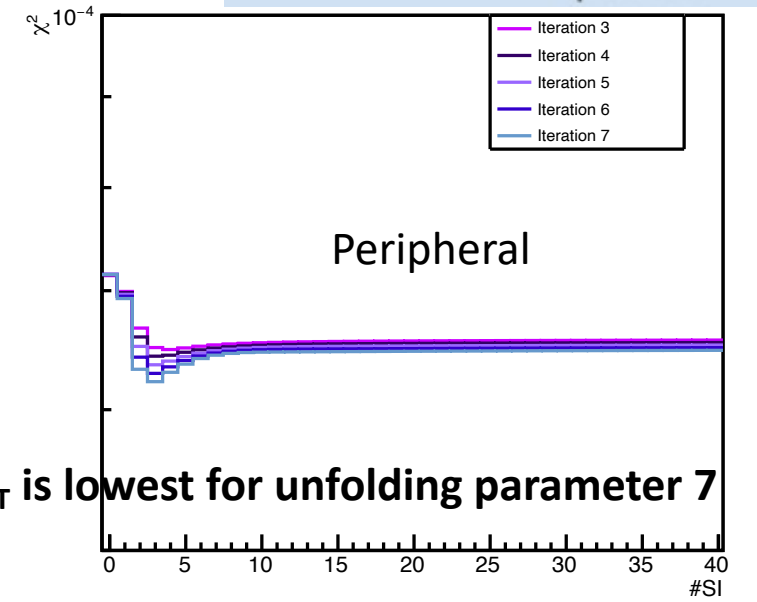
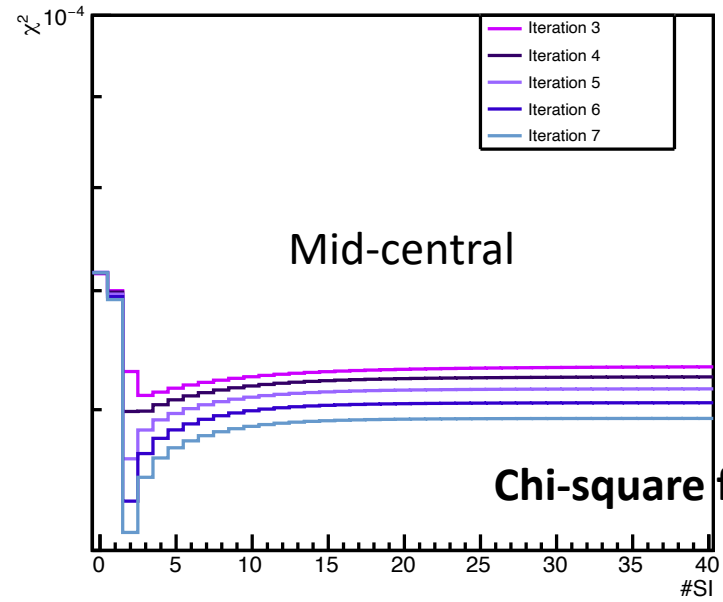
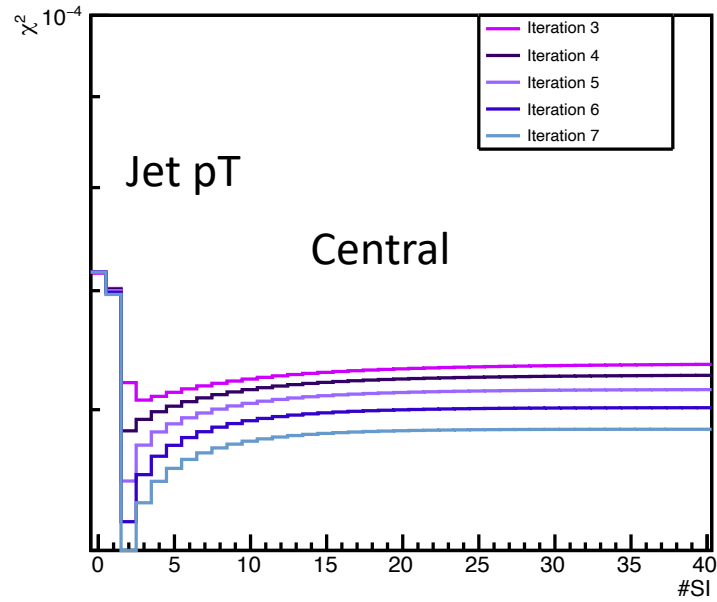
$$\chi^2 = \sum \frac{(\text{Observed} - \text{Expected})^2}{\text{Expected}}$$

For data, I use the PYTHIA truth as the expected value.

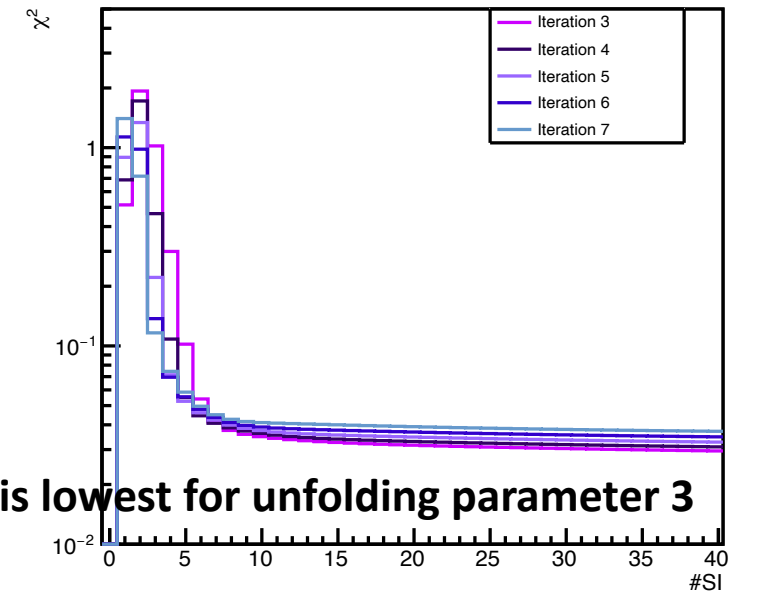
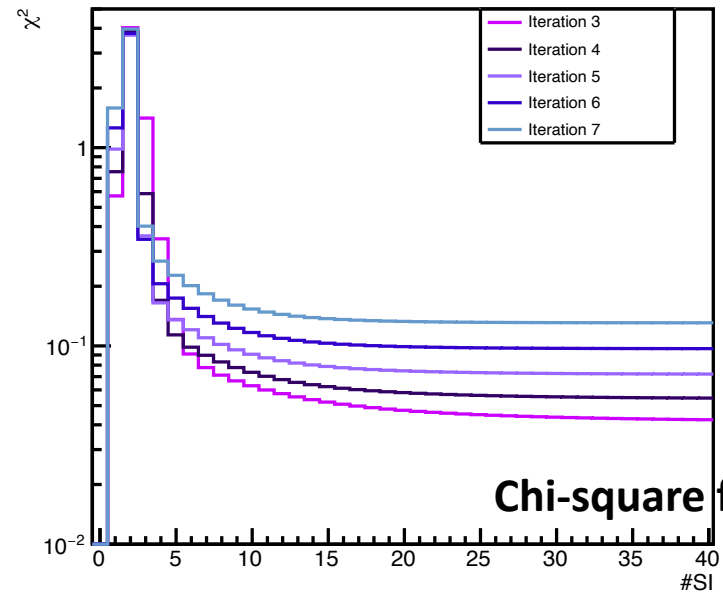
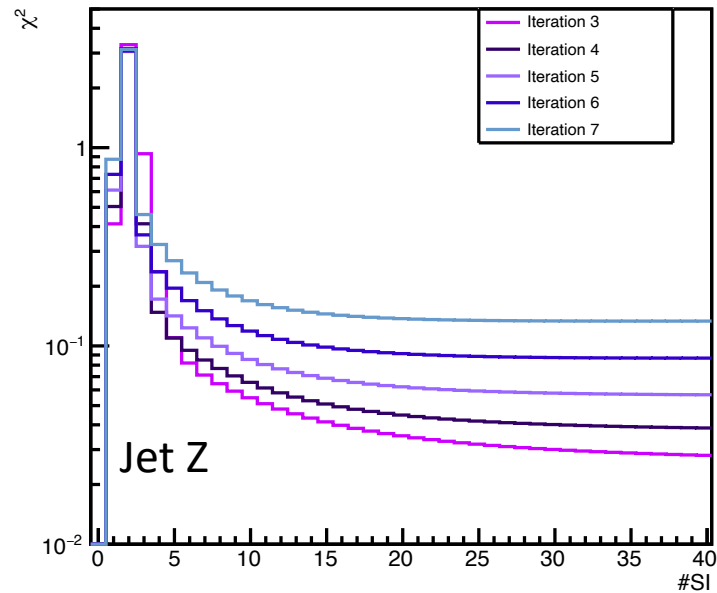
Chi-Square after superiterations:

Chi-Square (summed
over all bins):

$$\chi^2 = \sum \frac{(\text{Observed} - \text{Expected})^2}{\text{Expected}}$$

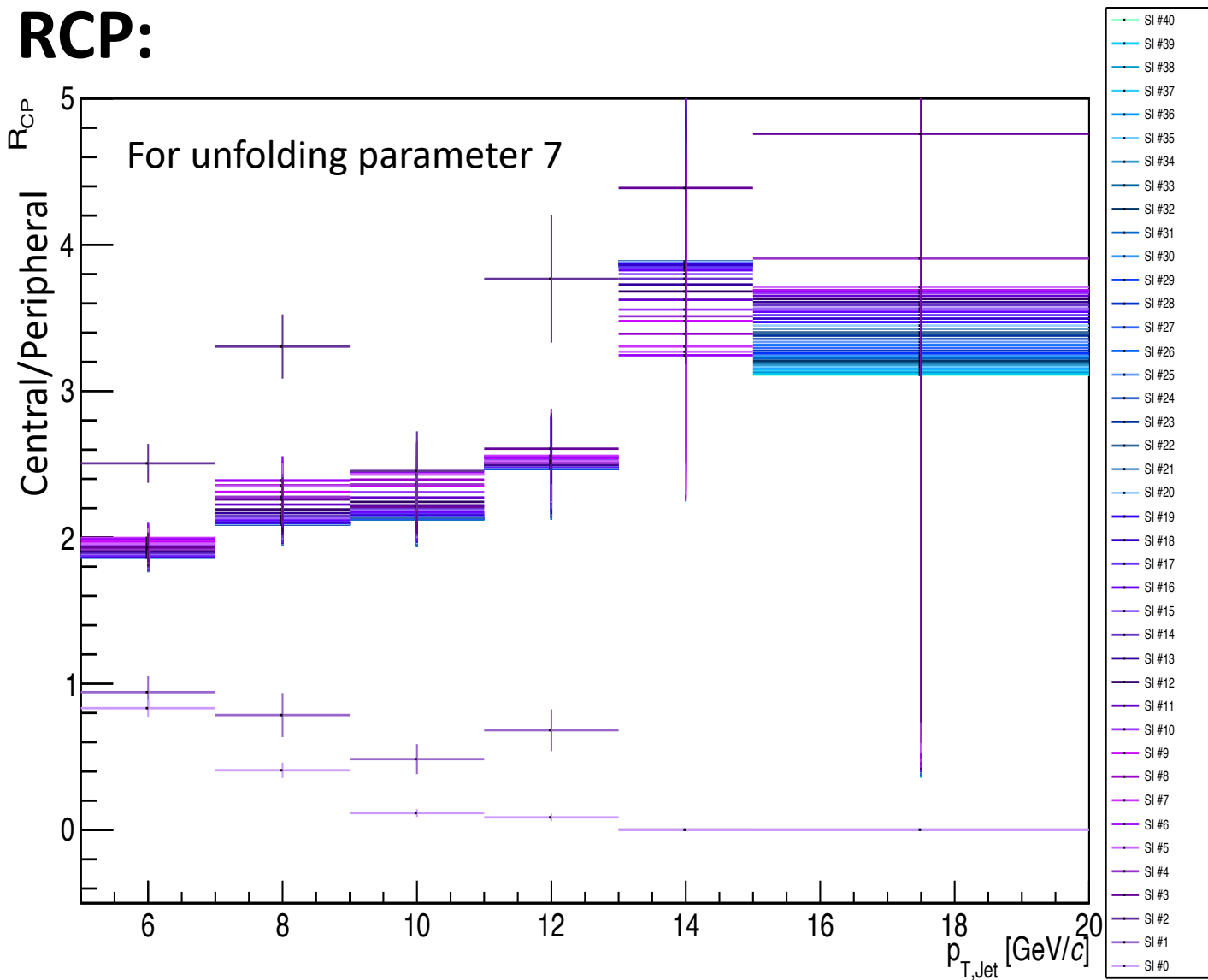


Chi-square for p_T is lowest for unfolding parameter 7



Chi-square for z is lowest for unfolding parameter 3

RCP:



- RCP stabilizes to values after ~ 10 Superiterations, same as chi2
- The RCP shows the correct trend for superiterations ~ 5 and above, but **it's greater than 2.**

- I checked all the normalisations we were doing earlier, and I am applying them correctly.
 - It still seems like I am missing some normalization, and I am stuck on what test I can run to figure this out.
- I checked RCP for self-similar and test-train samples with this method, and they are around 1 with some fluctuations ($\sim 20\%$).