

# Unfolding of Inclusive pAu P<sub>T</sub> Data

Slide 1

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[2019/08/01 @ 3 pm JST]  
**RadLab Meeting**

## CONTENT

Covariance (error) matrix for the unfolded spectrum using toy MC

⌘ Data covariance matrix used as input to generate toys – matrices compared for kregs.

Covariance (error) matrix on the unfolded spectrum from finite detector matrix statistics

⌘ Unfolded covariance matrix from response matrix with kreg comparisons

Cross-checks for unfolded spectrum

⌘ Cross-check using singular value distribution

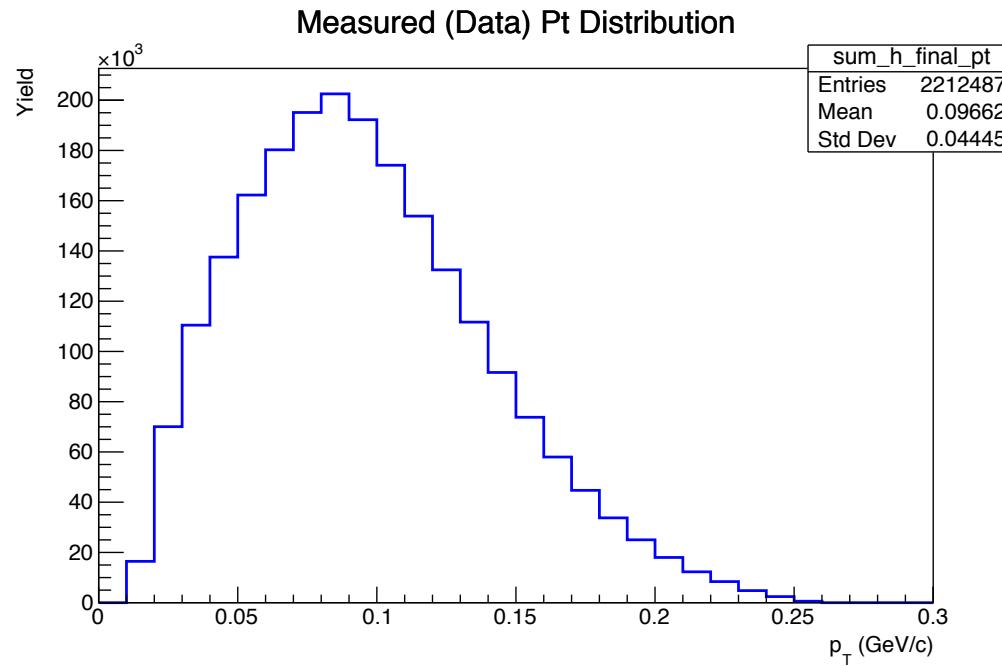
⌘ Trivial cross-check using UPC+DPMJET response matrix on UPC sample

# Covariance Matrix for Unfolded Spectrum from Measured Spectrum

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Three parameters required:

⌘ Covariance matrix on data



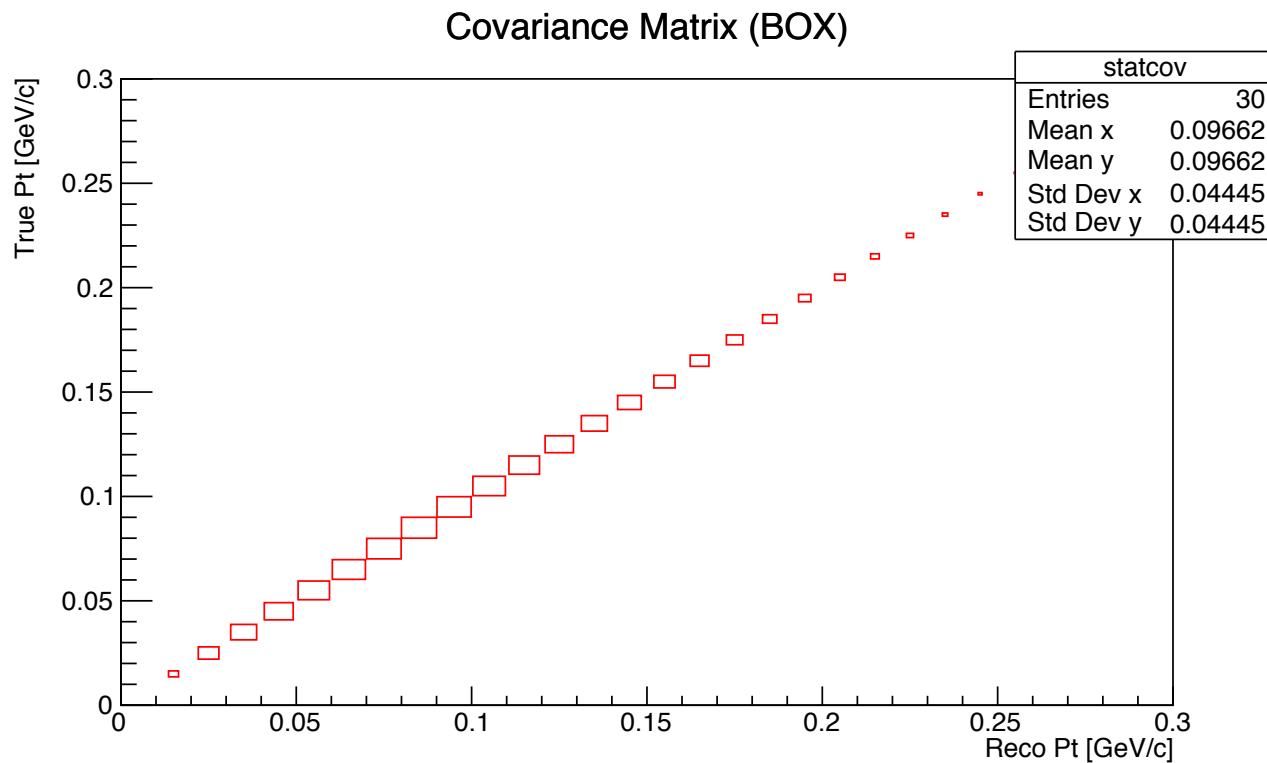
⌘ Pseudo experiments

⌘ Seed for pseudo experiments.

# Covariance Matrix for Unfolded Spectrum from Measured Spectrum

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- ◎ Diagonal covariance matrix from measured spectrum



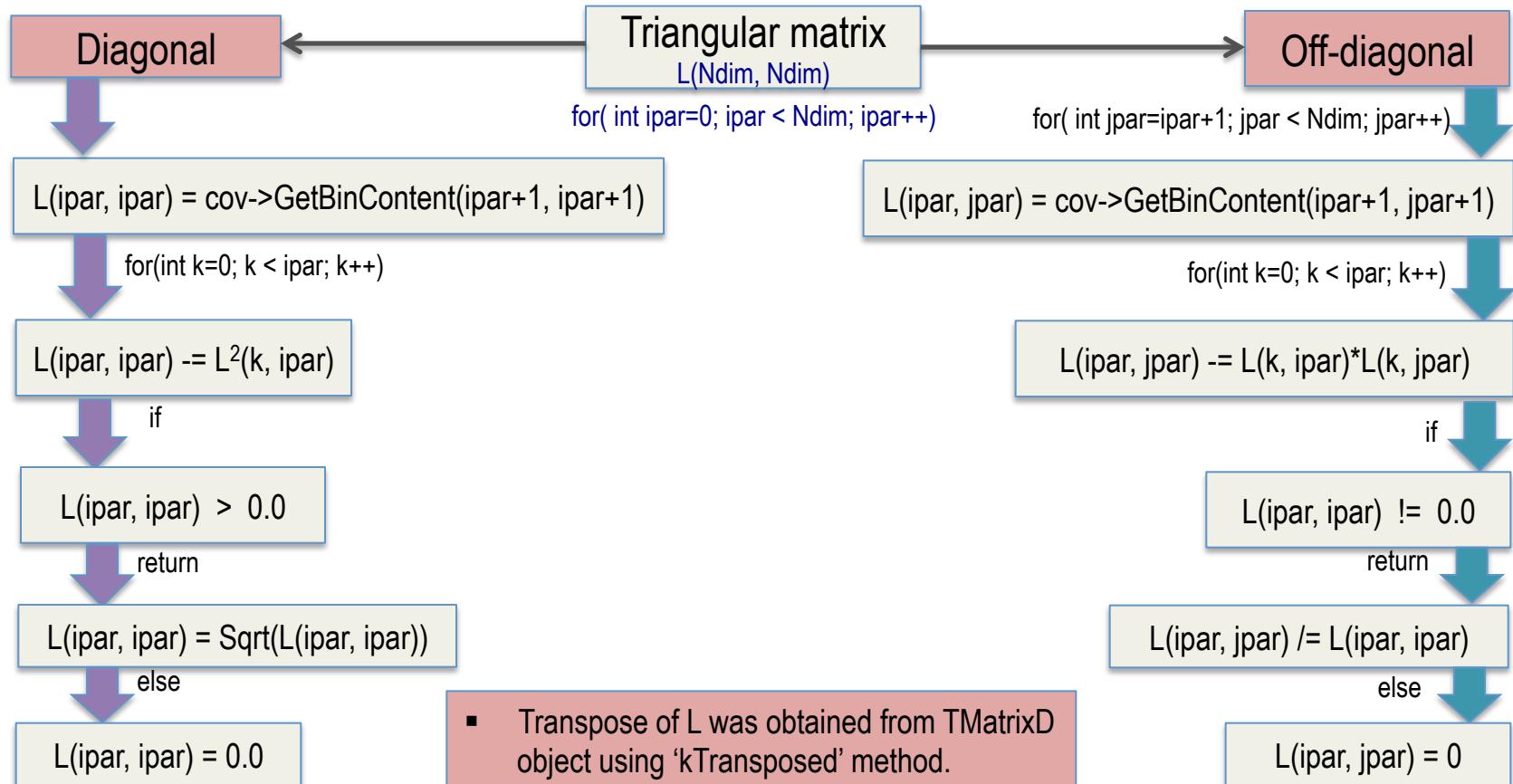
- ◎ This covariance matrix was then used to compute error matrix for the unfolded spectrum using toy MC.

# Covariance Matrix for Unfolded Spectrum from Measured Spectrum

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Toy generation:

- Elements of the upper-triangular matrix, L that give  $L^*L = C$  were calculated with  $L^t = \text{transpose of } L$ .  
Triangular matrix such that  $L = L(Ndim, Ndim)$  with  $Ndim = \text{dimension}$



# Covariance Matrix for Unfolded Spectrum from Measured Spectrum

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1D toy histogram is cloned from measured pt distribution (data). Toy mean is obtained by:

- Creating vector of unit Gaussian variables  $g(k) = \text{random.Gaus}(0., 1.)$
- Multiplying  $g(k)$  by  $L_t$ ,  $g(k) *= L_t$
- Adding mean value offsets

```
toyhistogram->SetBinContent( j, Bdata->GetBinContent(j) + g(j-1) )
```

```
toyhistogram->SetBinError( j, Bdata->GetBinError(j) )
```

Then performed unfolding with regularization parameter

- $K_{reg} = 4, 6, 8, 10$  with optimal  $k_{reg}$  value being 6.  

```
toymean ->SetBinContent( j, toymean->GetBinContent(j) + unfres->GetBinContent(j)/ntoys )
```

Reset random seed and deal with toys for the covariance matrix.

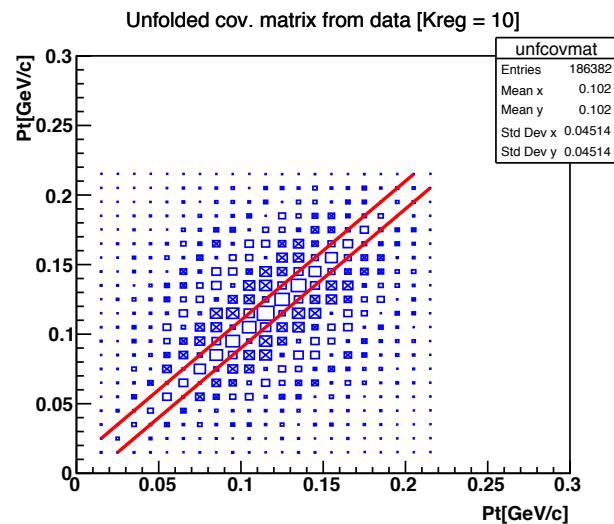
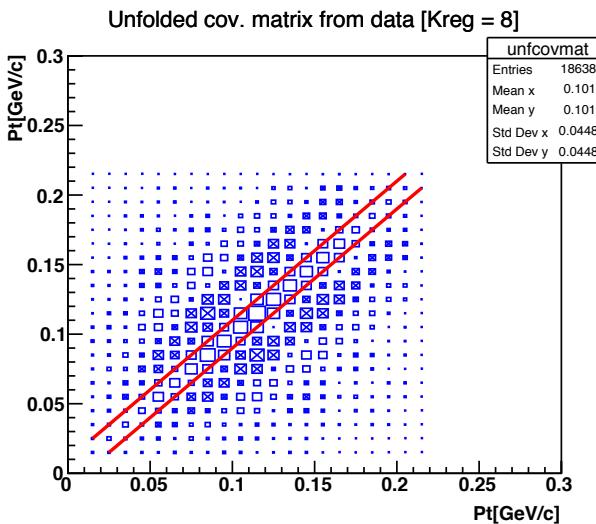
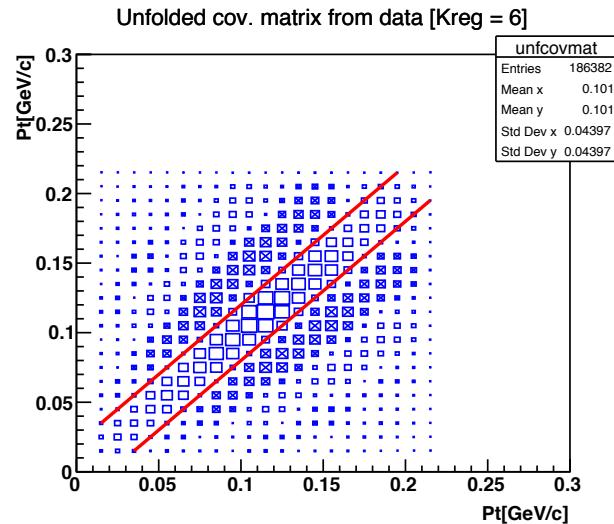
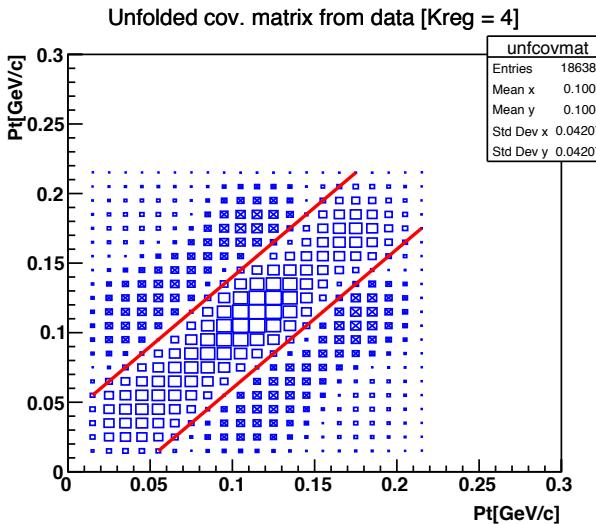
- Created vector of unit Gaussian variables  $g(k) = \text{random.Gaus}(0., 1.)$
- Multiplied  $g(k)$  by  $L_t$ ,  $g(k) *= L_t$
- Added mean value offsets

Unfolded with regularization parameter from dvector and stored the unfolded covariance matrix.

- Used  $k_{reg} = 4, 6, 8$  and 10 for comparison purpose.
- ```
UnfCov->SetBinContent( j, k, UnfCov->GetBinContent(j,k) + ( (unfres->GetBinContent(j) - toymean->GetBinContent(j) ) * (unfres->GetBinContent(k) - toymean->GetBinContent(k))/(ntoys-1) ) )
```

# Covariance Matrix for Unfolded Spectrum from Measured Spectrum

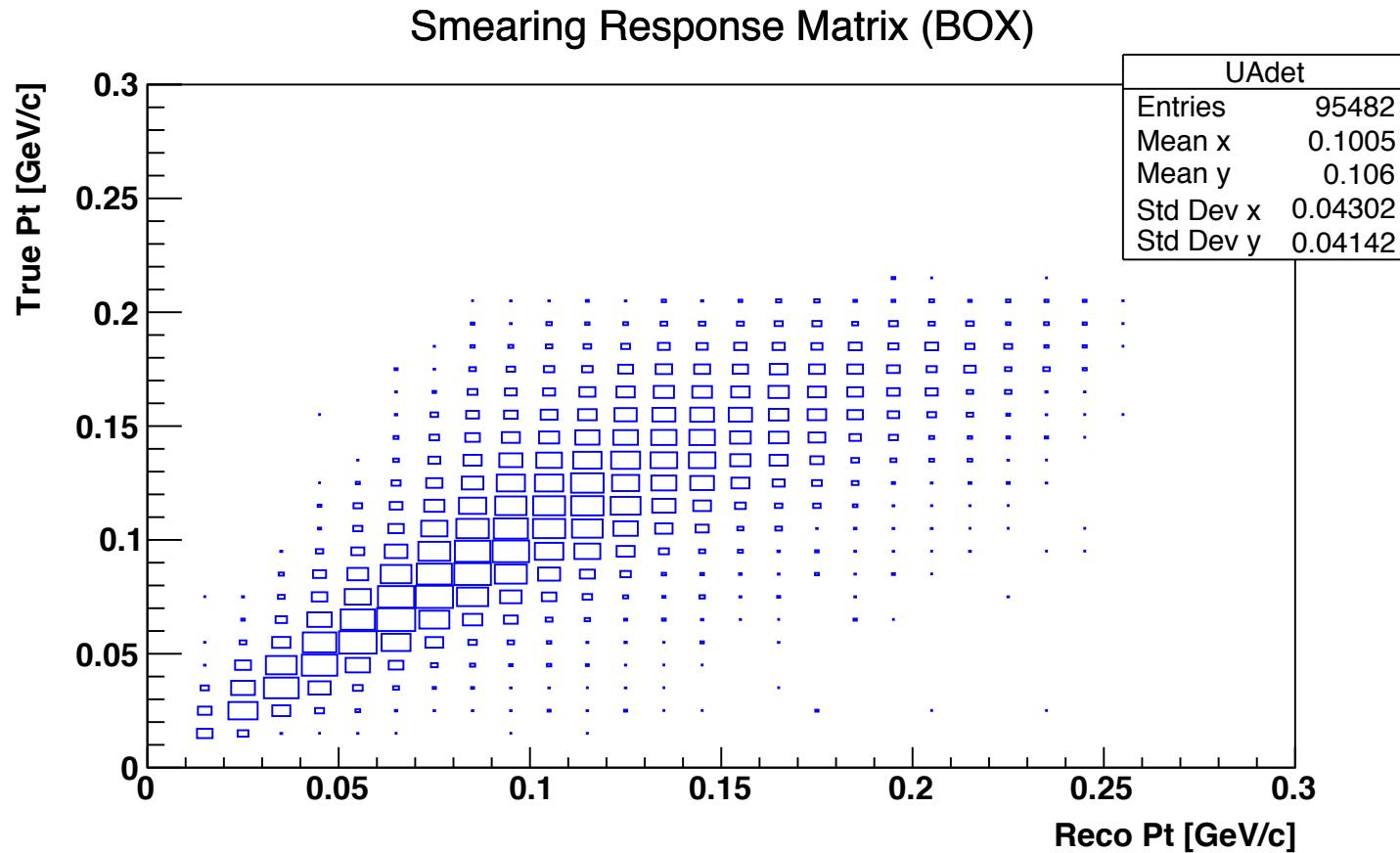
Slide 6



# Covariance Matrix for Unfolded Spectrum from Response Matrix

Slide 7

Detector response matrix: 2D histogram extracted from the reco and true pt spectra of UPC+DPMJET MC



# Covariance Matrix for Unfolded Spectrum from Response Matrix

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Two arguments required: (1) ntoys and (2) random seed for pseudo experiments

- A copy of response matrix is made for unfolded covariance matrix.

Toys for detector response matrix are generated using a TRandom3 generator with seed

- Toy matrix (toymat) is cloned from detector response matrix
- Toy mean 1D histogram is also cloned from true MC distribution.
- Toy matrix is then generated based on toy number and dimension of response matrix using Poisson.  
`toymat->SetBinContent(k, m, random. Poisson(ResMat->GetBinContent(k,m)) )`

Unfolding with appropriate regularization parameter (kreg) is performed

- Kreg = 4, 6 (optimal), 8 and 10 plotted for comparison
- Toy mean (toymean) is stored as:  
`toymean->SetBinContent( j, toymean->GetBinContent(j) + unfres->GetBinContent(j)/ntoys`

Random seed is reset for unfolded covariance matrix

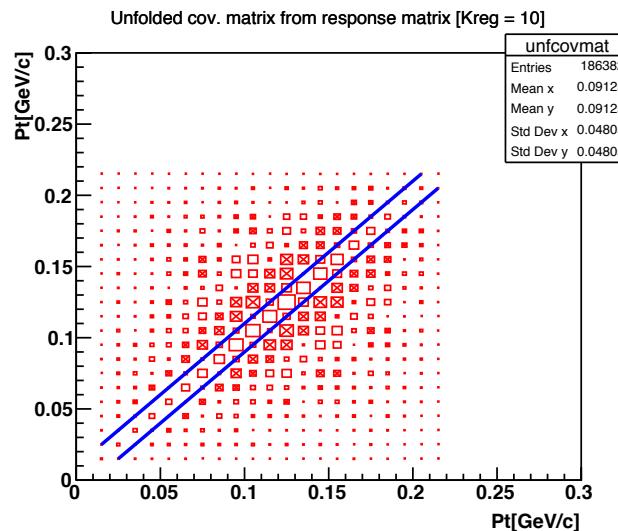
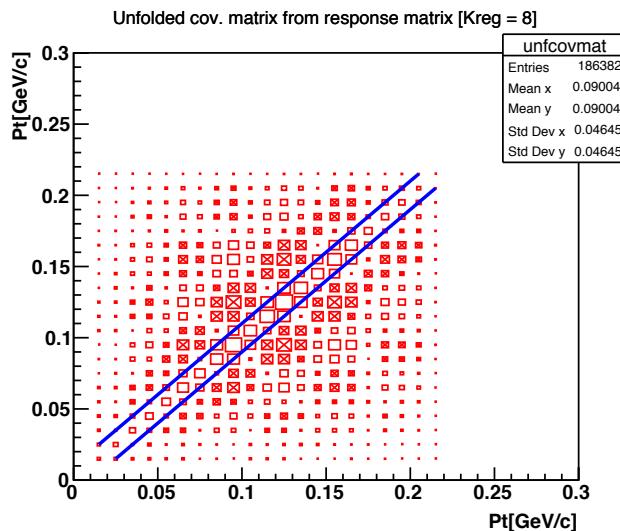
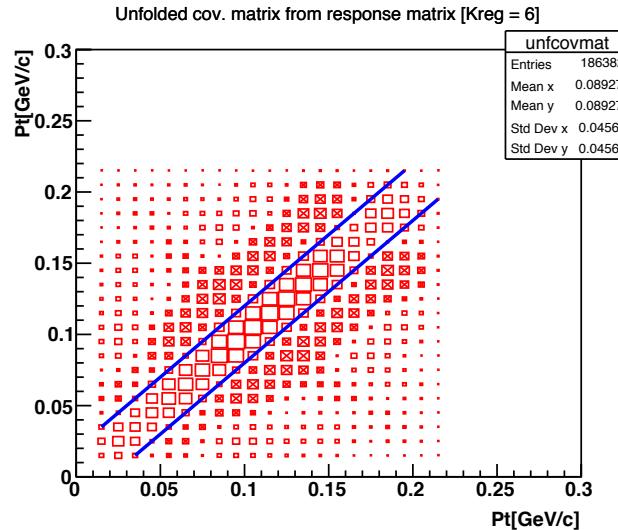
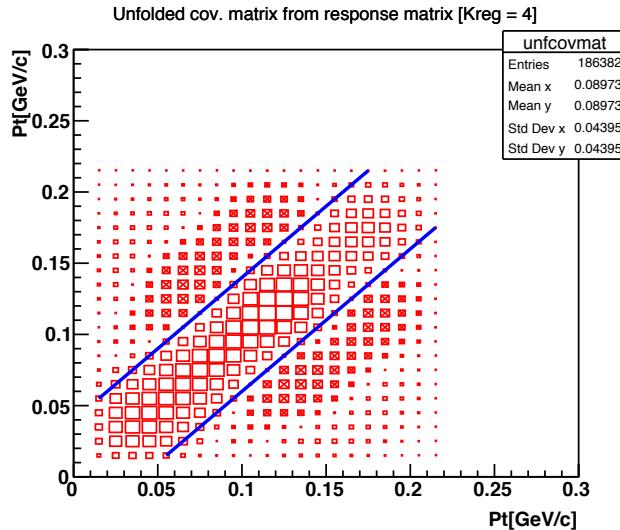
- Toy matrix is generated based on toy number and dimension of response matrix using Poisson  
`toymat->SetBinContent(k, m, random. Poisson(ResMat->GetBinContent(k,m)) )`

Unfolding is performed with regularization parameter = 4, 6, 8, 10 and unfolded covariance matrix is stored.

- `Unfcov->SetBinContent(j, k, unfcov->GetBinContent(j, k) + (( unfres->GetBinContent(j) – toymean->GetBinContent(j) ) * (unfres->GetBinContent(k) – toymean->GetBinContent(k) )/(ntoys-1) )`

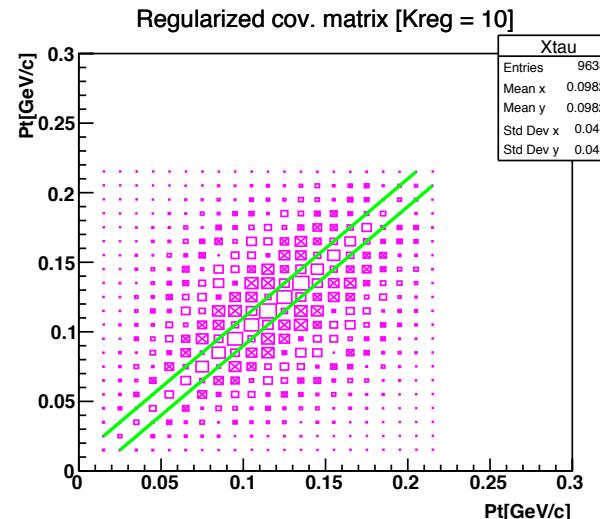
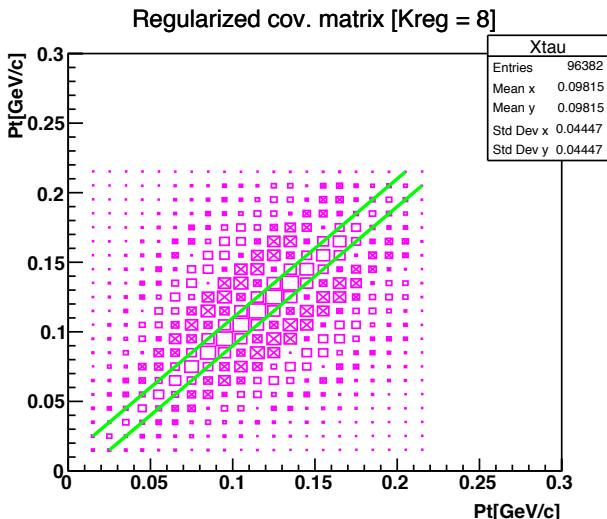
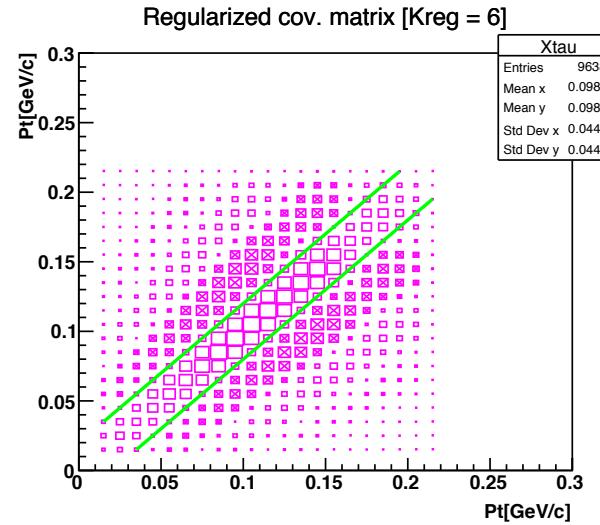
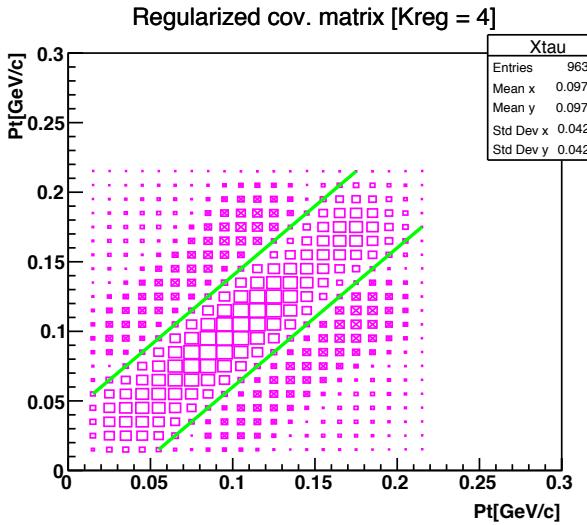
# Covariance Matrix for Unfolded Spectrum from Response Matrix

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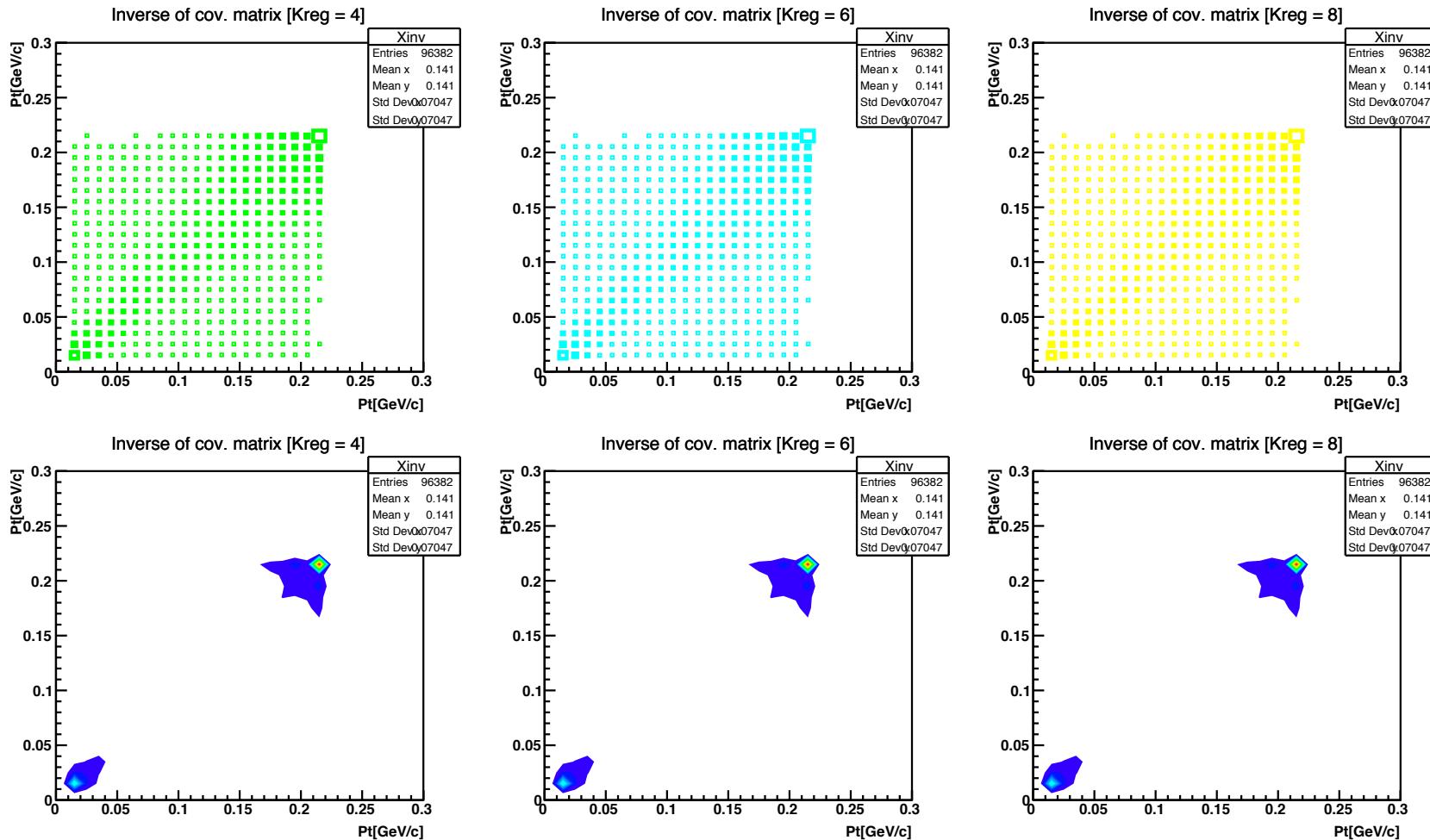
# Regularized Covariance Matrix Corresponding to Total Uncertainties on Data

Slide 10



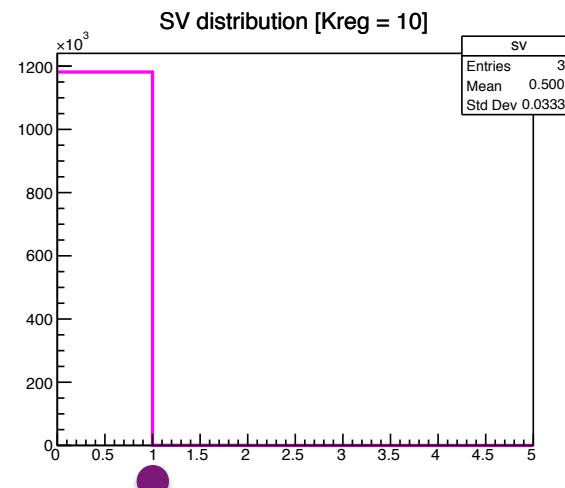
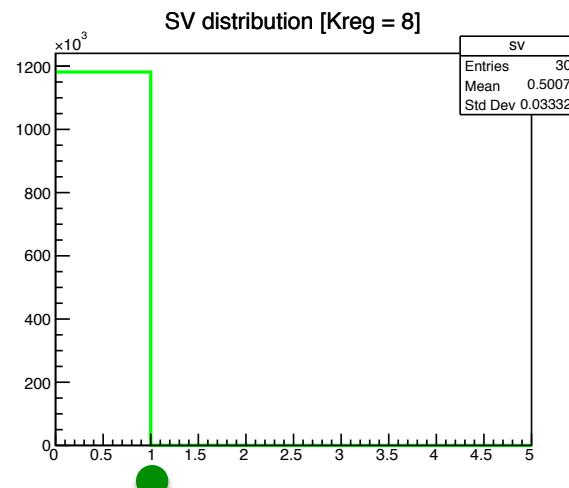
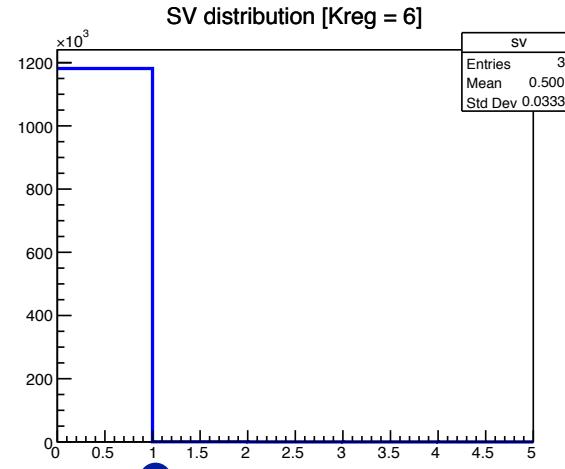
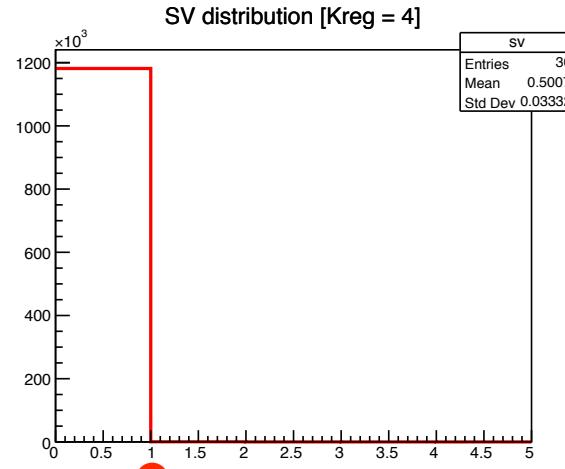
# Inverse of Covariance Matrix – Regularization Independent

Slide 11



# Singular Value Distributions – Unfolding Cross-check

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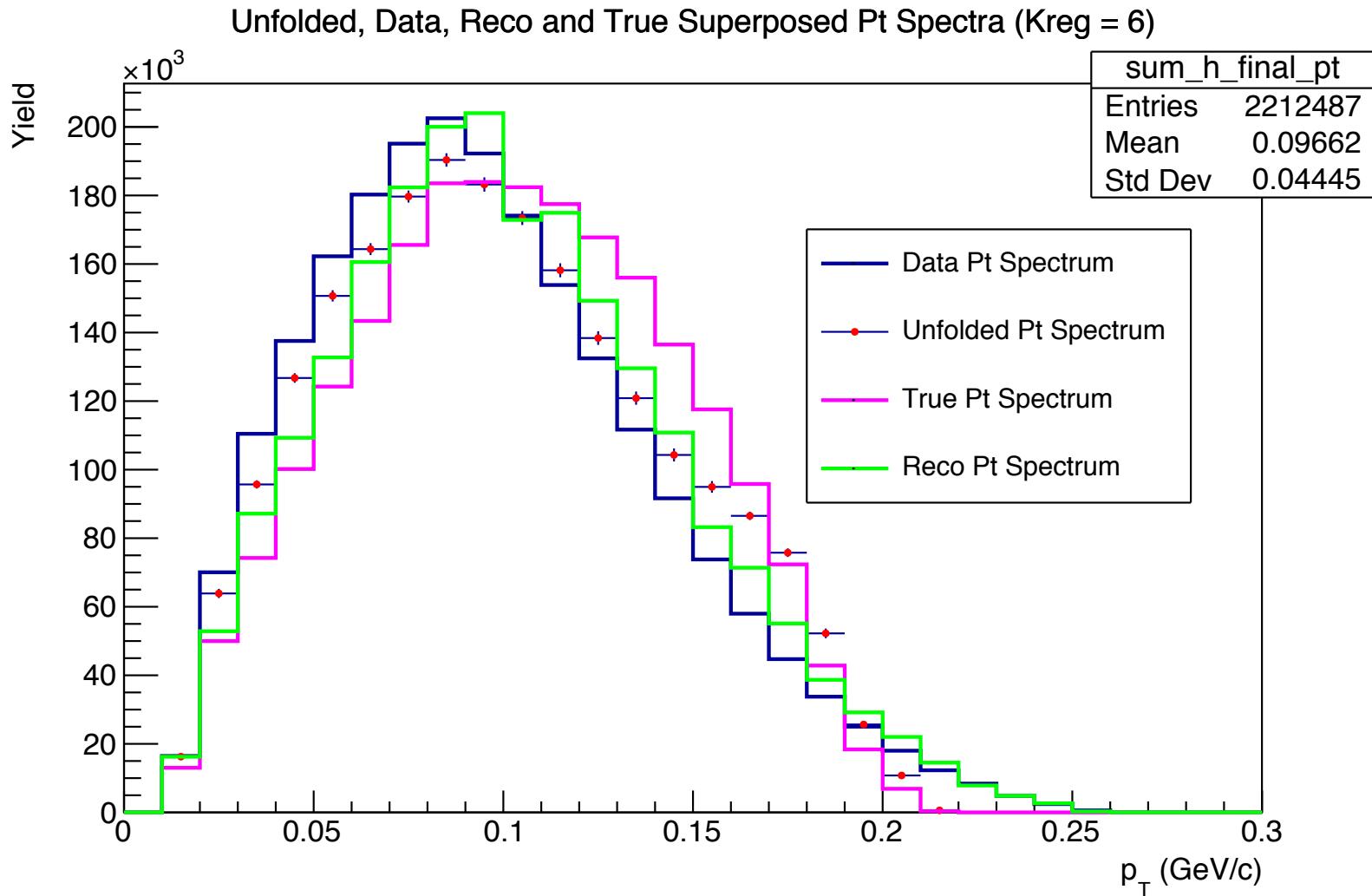


Singular values are distributed as expected and are regularization independent.

# Trivial Unfolding Cross-check – UPC+DPMJET

## Response Matrix on UPC+DPMJET

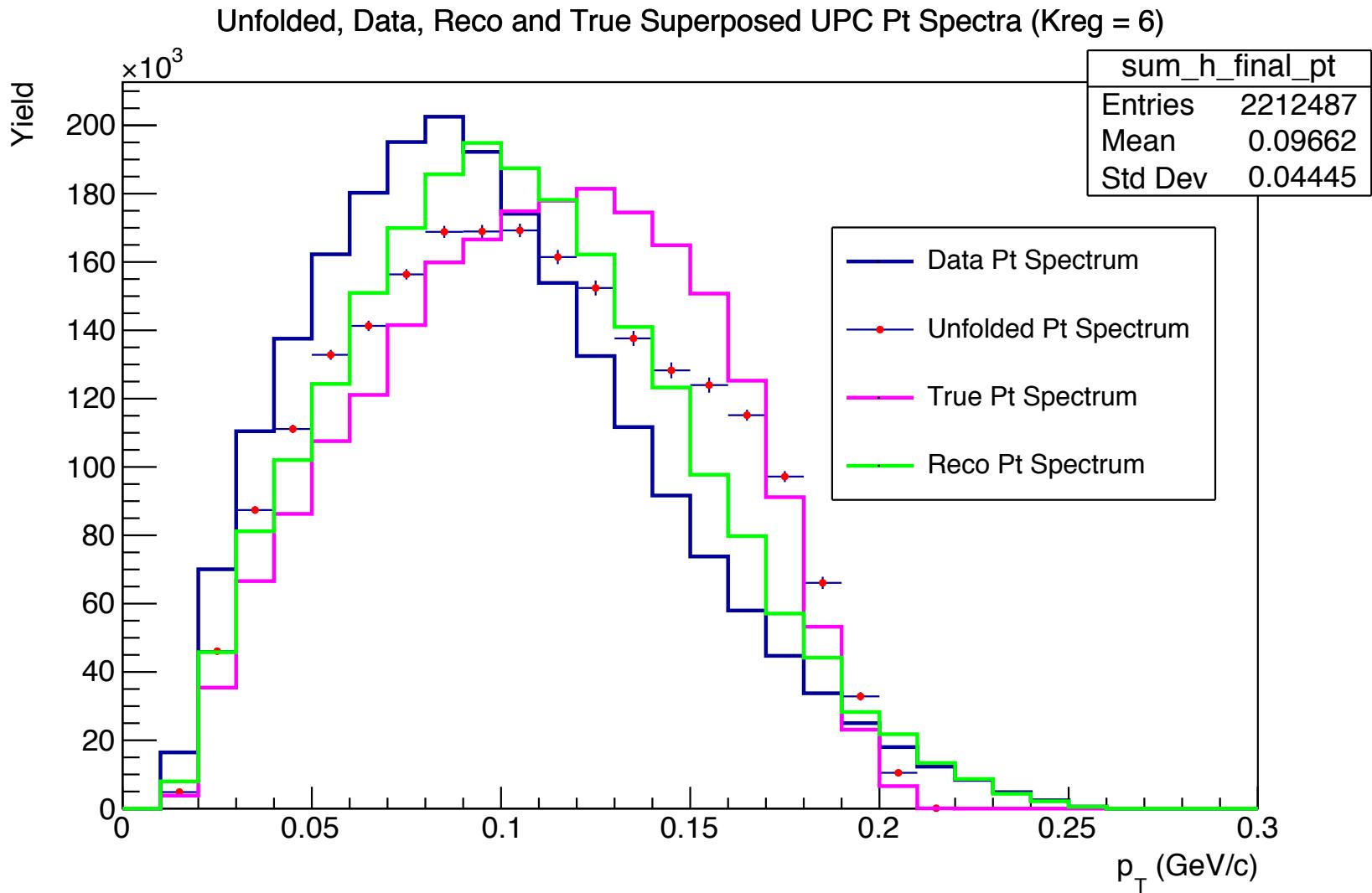
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# Trivial Unfolding Cross-check – UPC+DPMJET

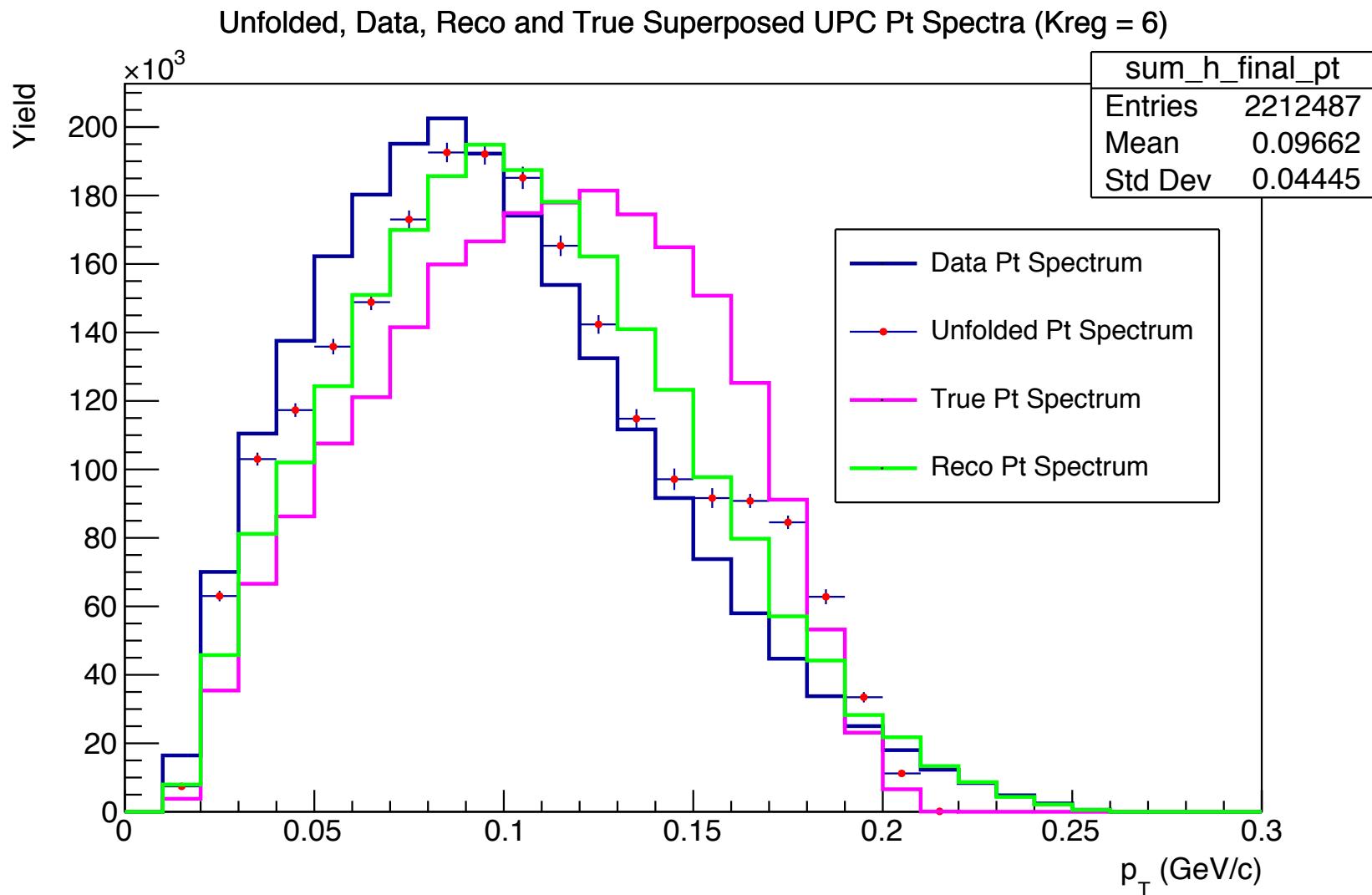
## Response Matrix on UPC

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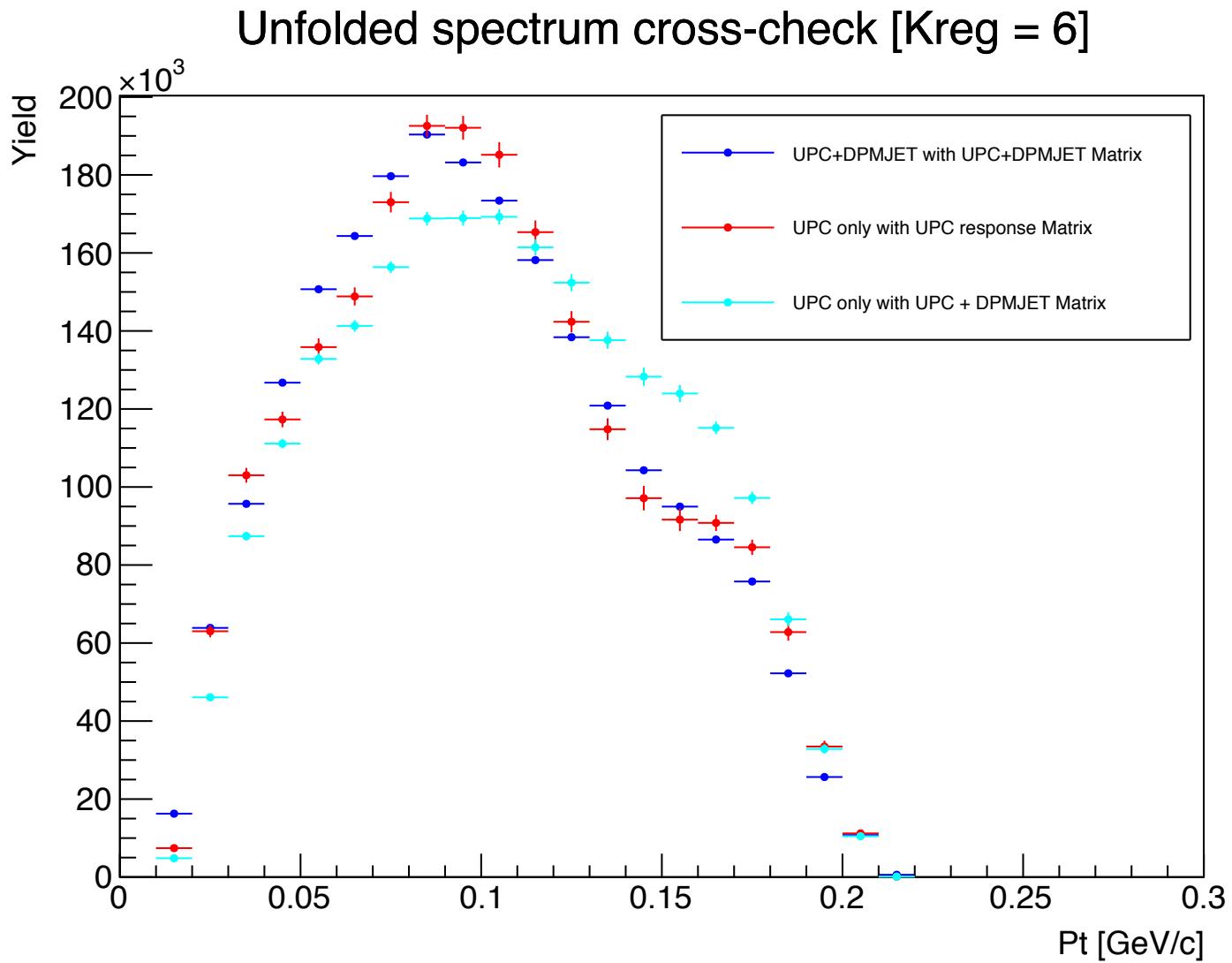
# Trivial Unfolding Cross-check – UPC Response Matrix on UPC

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# Trivial Unfolding Cross-check – Unfolded Spectra Comparison

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# **Next Task**

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- ⌘ Way forward to be decided depending on the outcome of this update

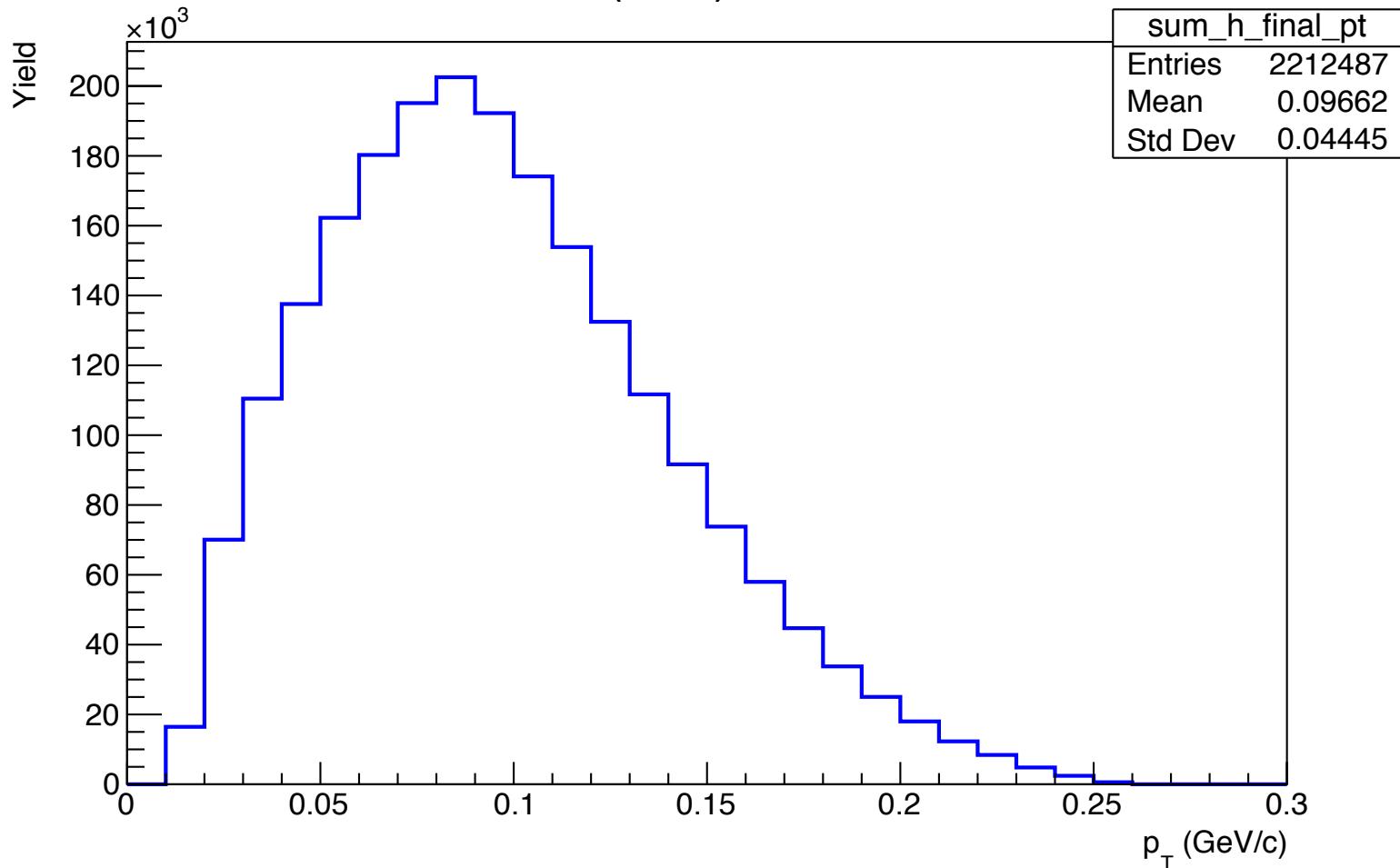
# **BACKUP**

# Unfolding Input – Measured Spectrum

Slide 1

- ① Data to unfold: Run 15 inclusive pAu transverse momentum data at  $\sqrt{s_{NN}} = 200$  GeV

Measured (Data) Pt Distribution



# Unfolding Input – Reco Spectrum

Slide 2

- ② Reconstructed spectrum: pAu reconstructed pt spectrum from a combination of UPC (EM) + DPMJET (HAD) MC samples.

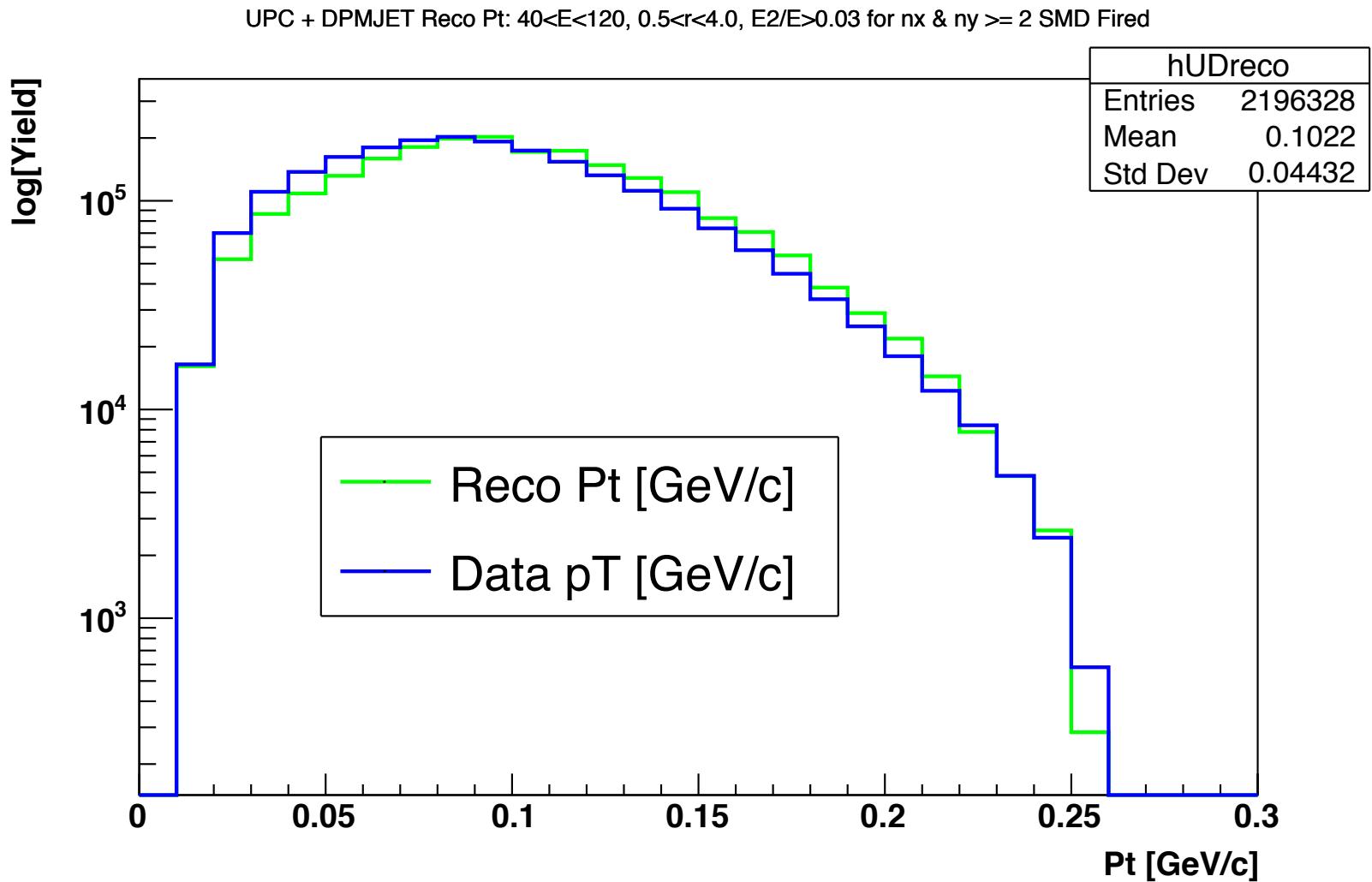
## Neutron Selection Cuts

Following cuts were utilized for neutron identification and rejection of photon events. Same cuts as data.

- ◎ ZDC total energy:  $40 < E < 120$  and  $2^{\text{nd}}$  ZDC energy/ZDC total energy  $> 0.03$  ( i.e. non-zero  $2^{\text{nd}}$  ZDC energy )
- ◎ Acceptance cut:  $0.5 < r < 4.0$  cm
- ◎ SMD multiplicity:  $N_x/N_y \geq 2$  fired SMD strips. That is  $N_x$  and  $N_y > 1$  fired strips above Minimum Ionized Particle (MIP) energy threshold cut.

# Unfolding Input – Reco Spectrum

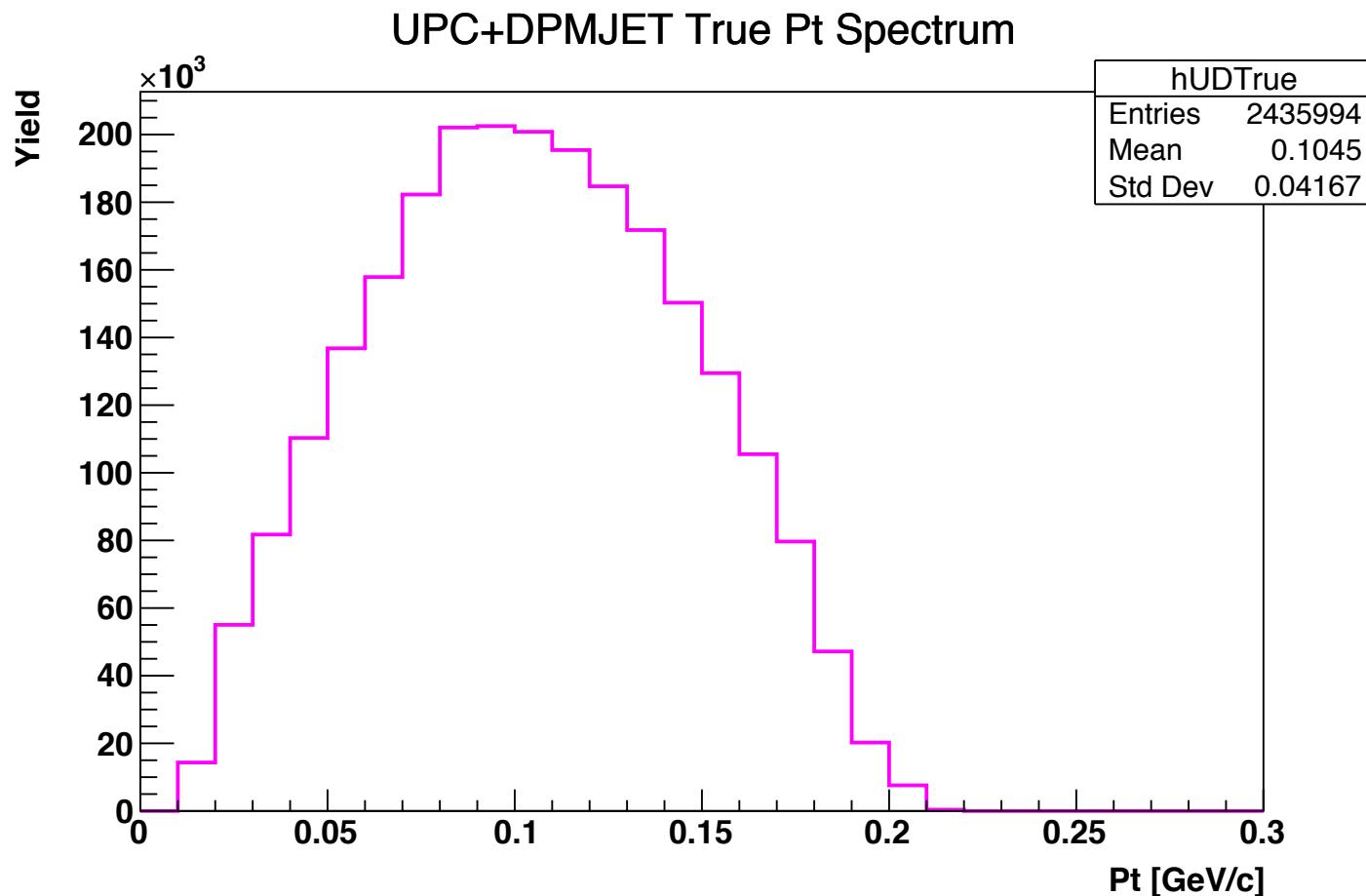
Slide 3



# Unfolding Input – True Spectrum

Slide 4

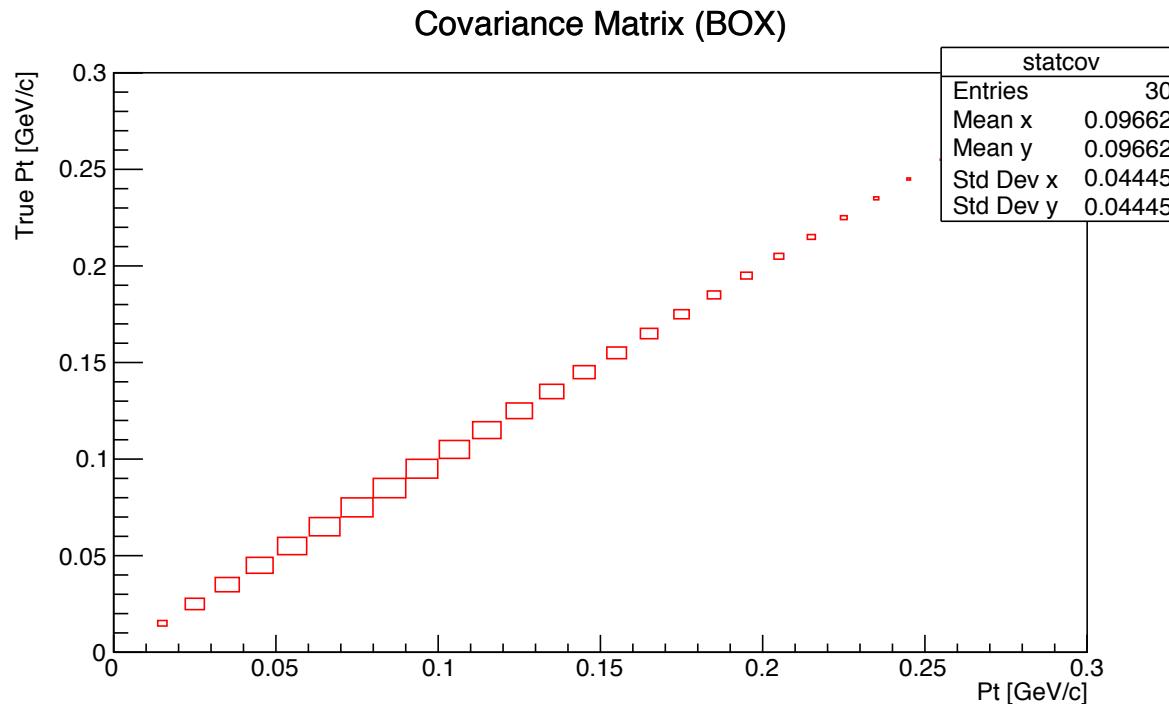
- ③ True spectrum: True pt spectrum from addition of UPC (EM) + DPMJET (HAD) MC samples.



# Unfolding Input – Covariance Matrix: BOX

Slide 5

- ⑤ Covariance matrix: 2D histogram extracted from the measured pt spectrum.



Finally created a TSVDUnfold object to perform unfolding of the measured spectrum

```
TSVDUnfold *tsvdunf = new TSVDUnfold( hdata, statcov, hUDreco, hUDTrue, UAdet )
```

Ready

1

2

3

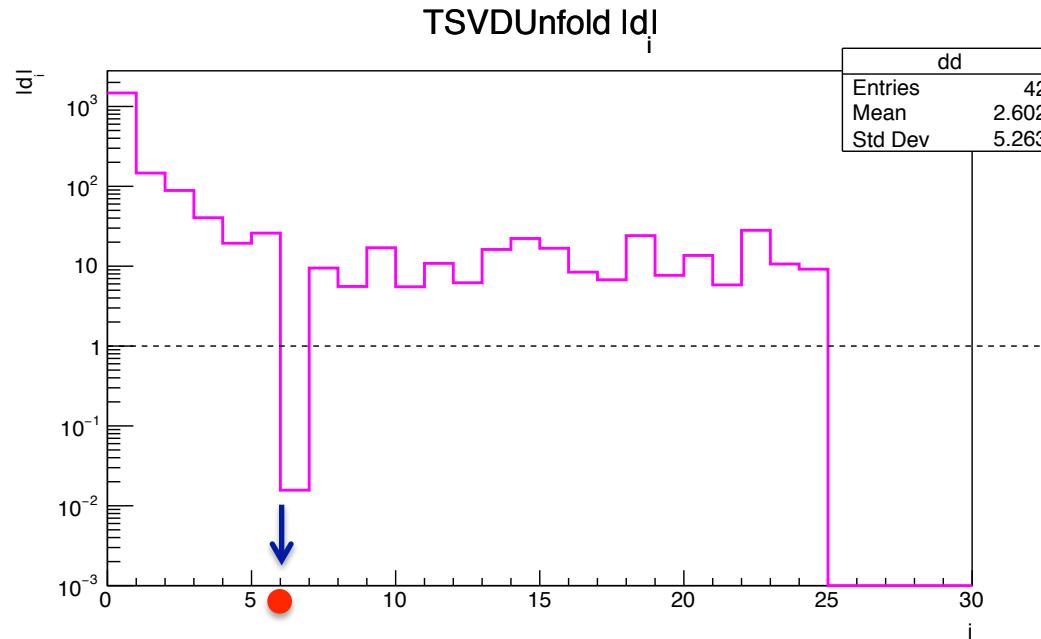
4

5

# Unfolding Output – Distribution of dVector

Slide 6

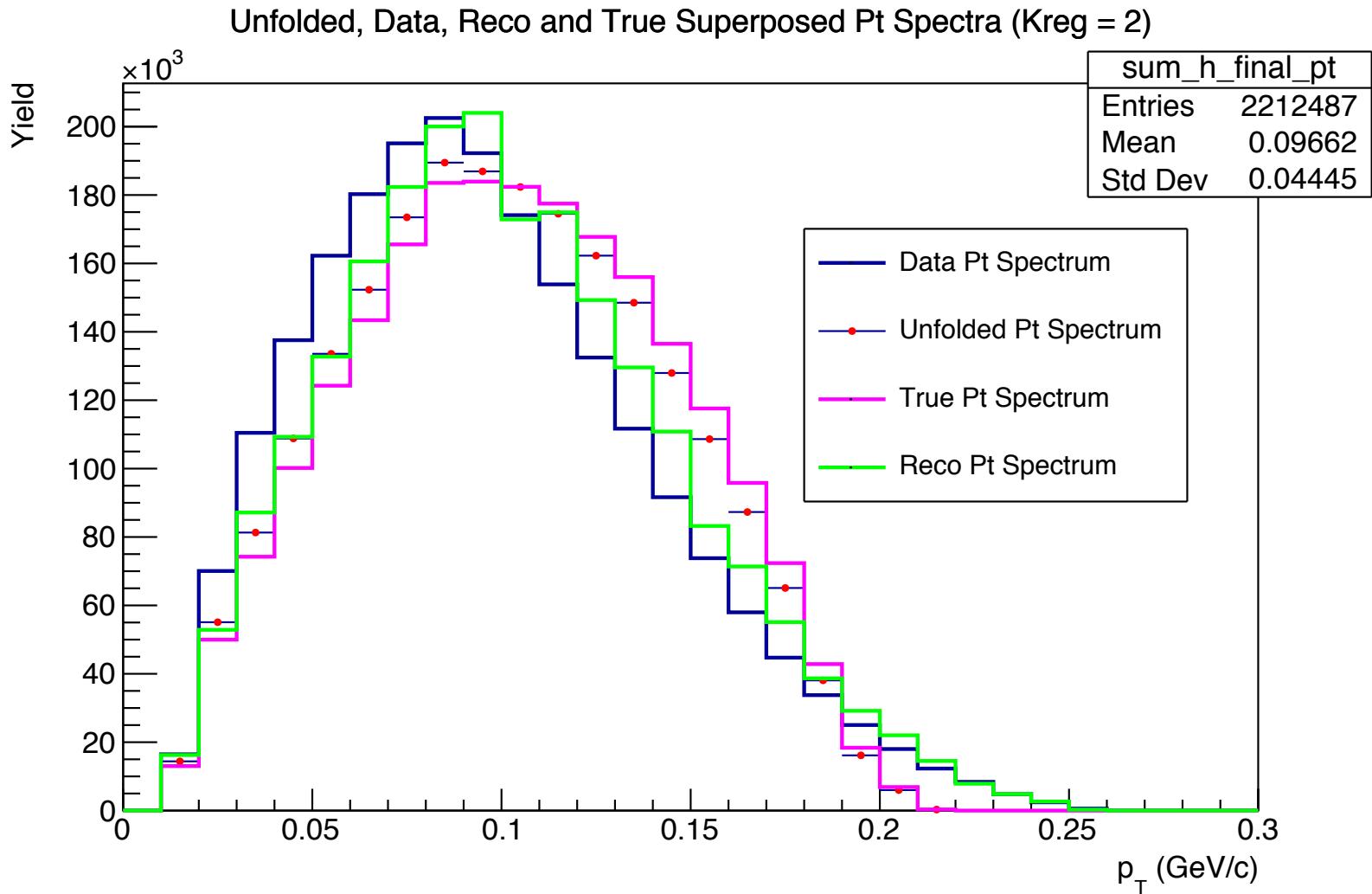
- ④ Performed unfolding with the regularization parameter ( $k_{reg} = 6$  as optimum).



- ④ This distribution helps us cross-check the quality of our unfolding regularization.
- ④ Regularization is chosen as the point where  $|d_i|$  stop being statistically significantly  $\gg 1$ . It is the point where regularization is considered optimal.

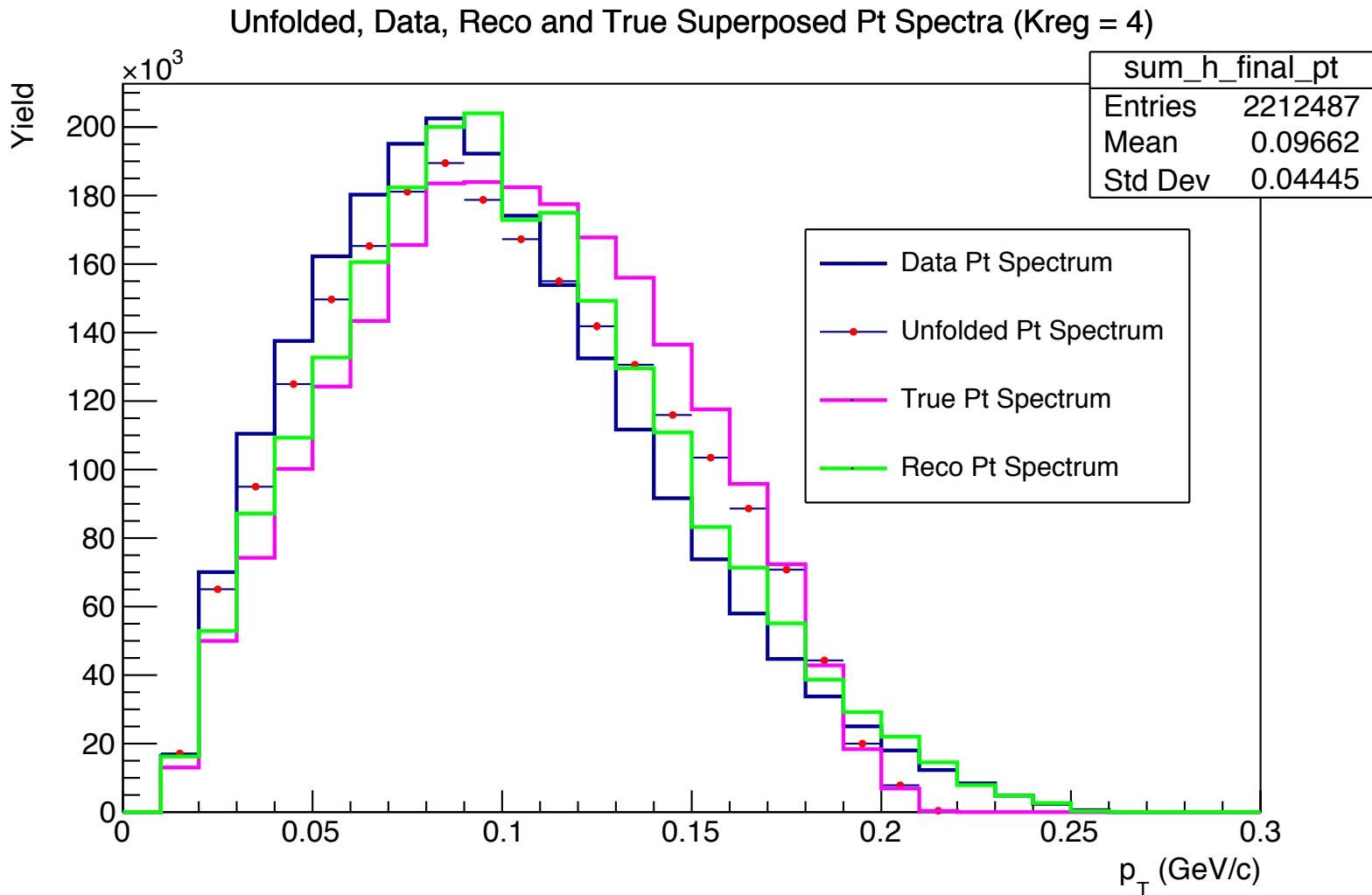
# Unfolding Output – Unfolded Distribution

Slide 7



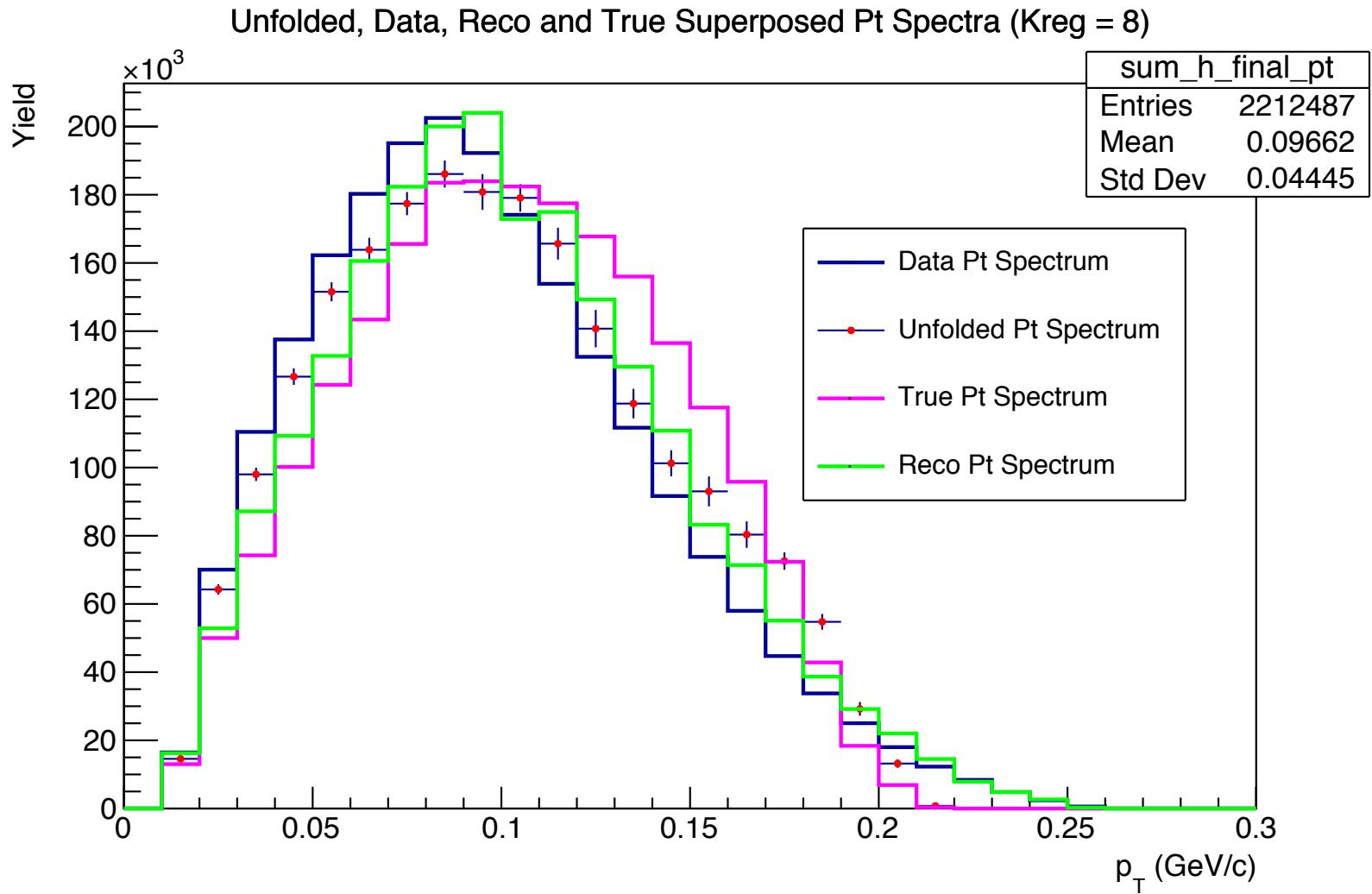
# Unfolding Output – Unfolded Distribution

Slide 8



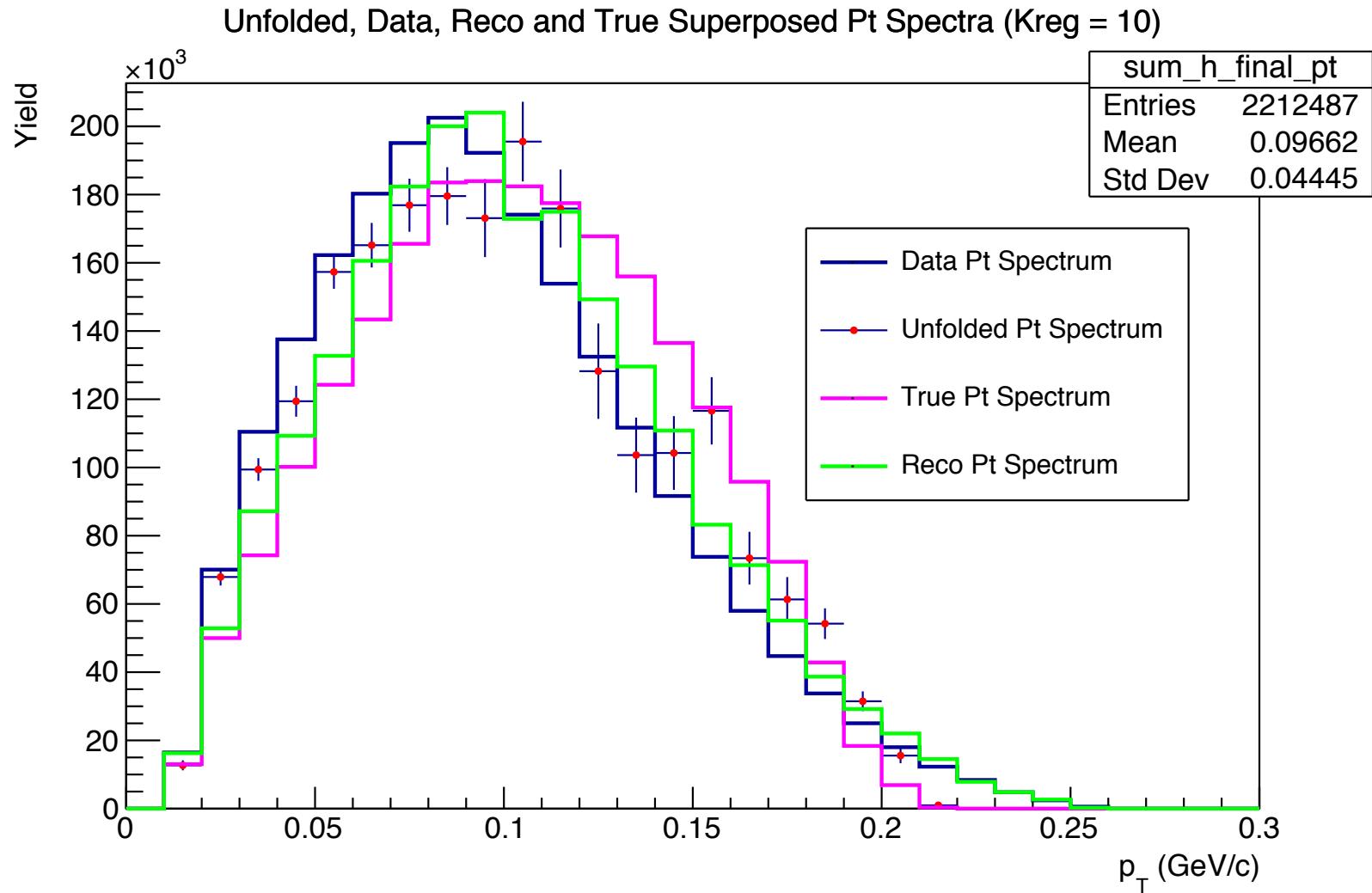
# Unfolding Output – Unfolded Distribution

Slide 9



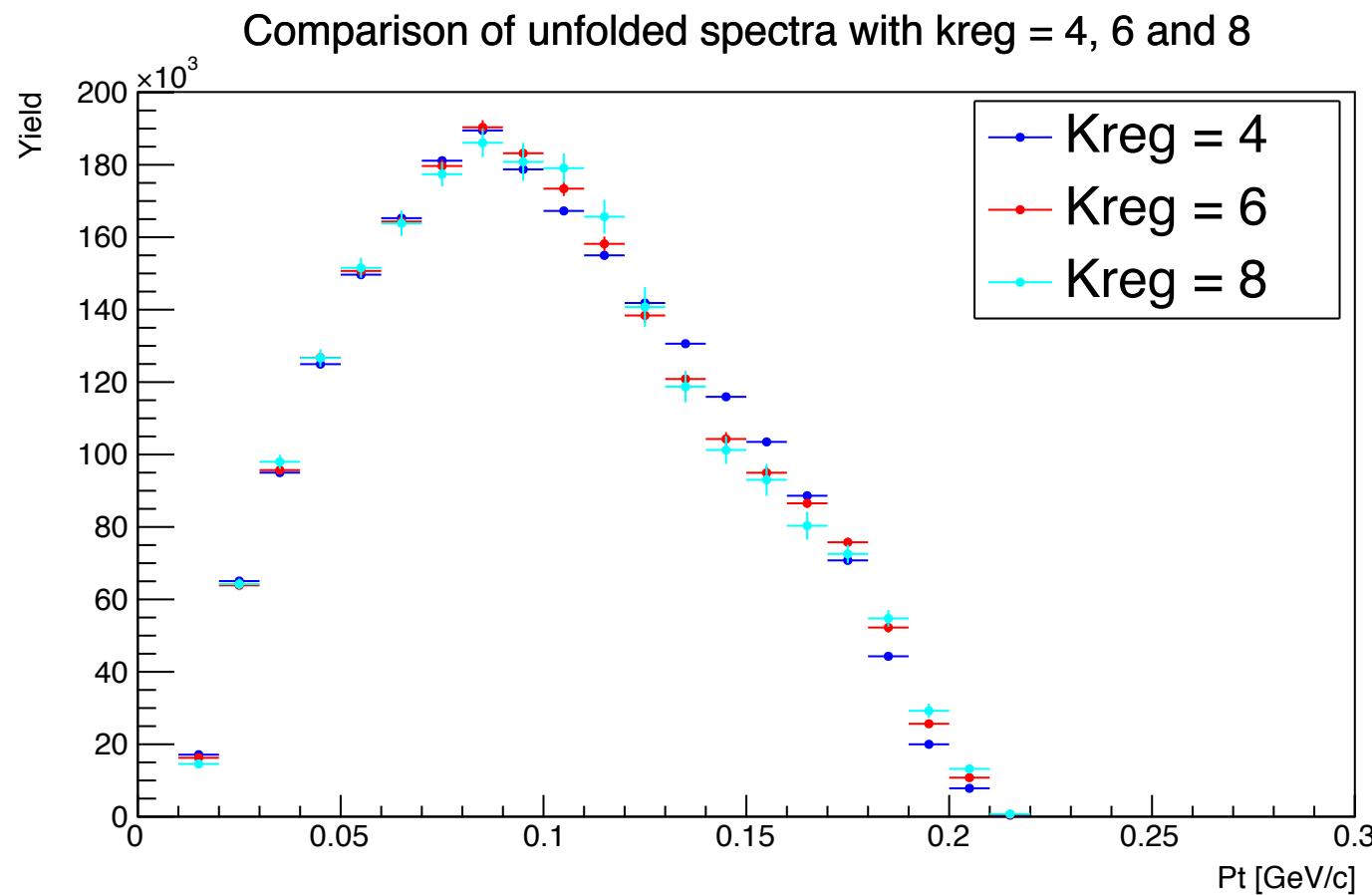
# BACKUP – Unfolded Distribution

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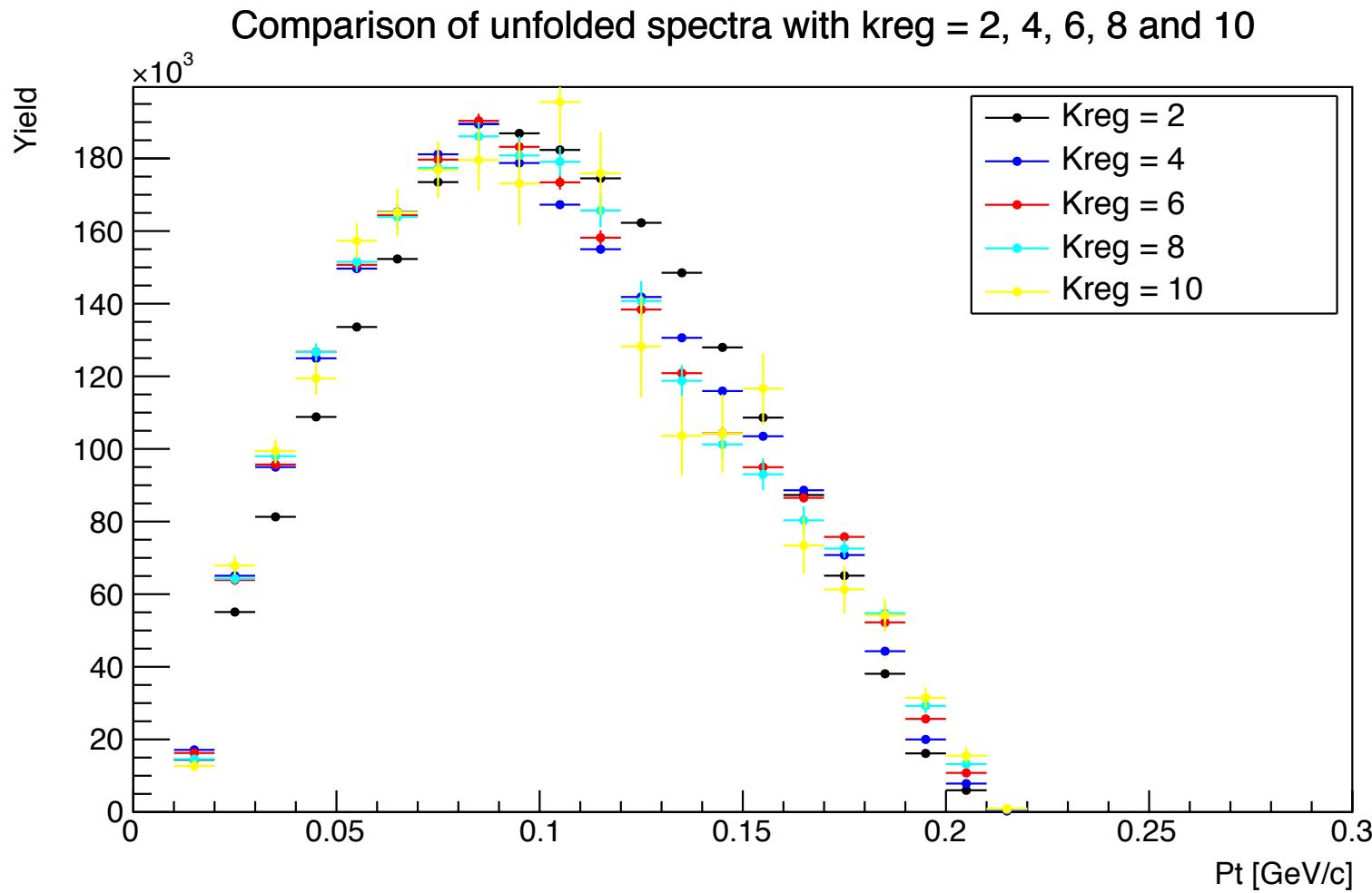
# Unfolded Inclusive pAu $P_T$ Data – Spin PWG

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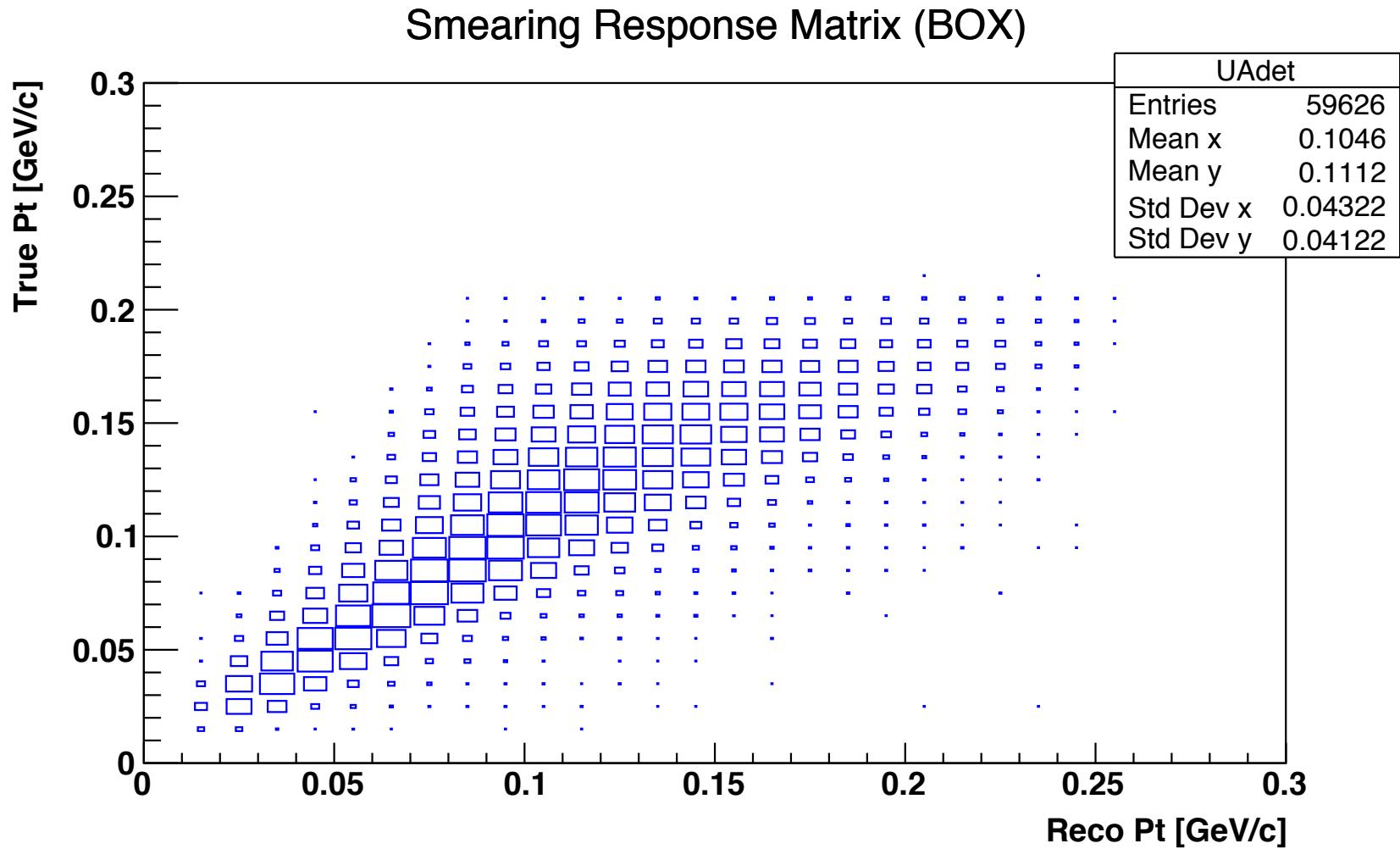
# Unfolding Output – Unfolded Distribution

Slide 12



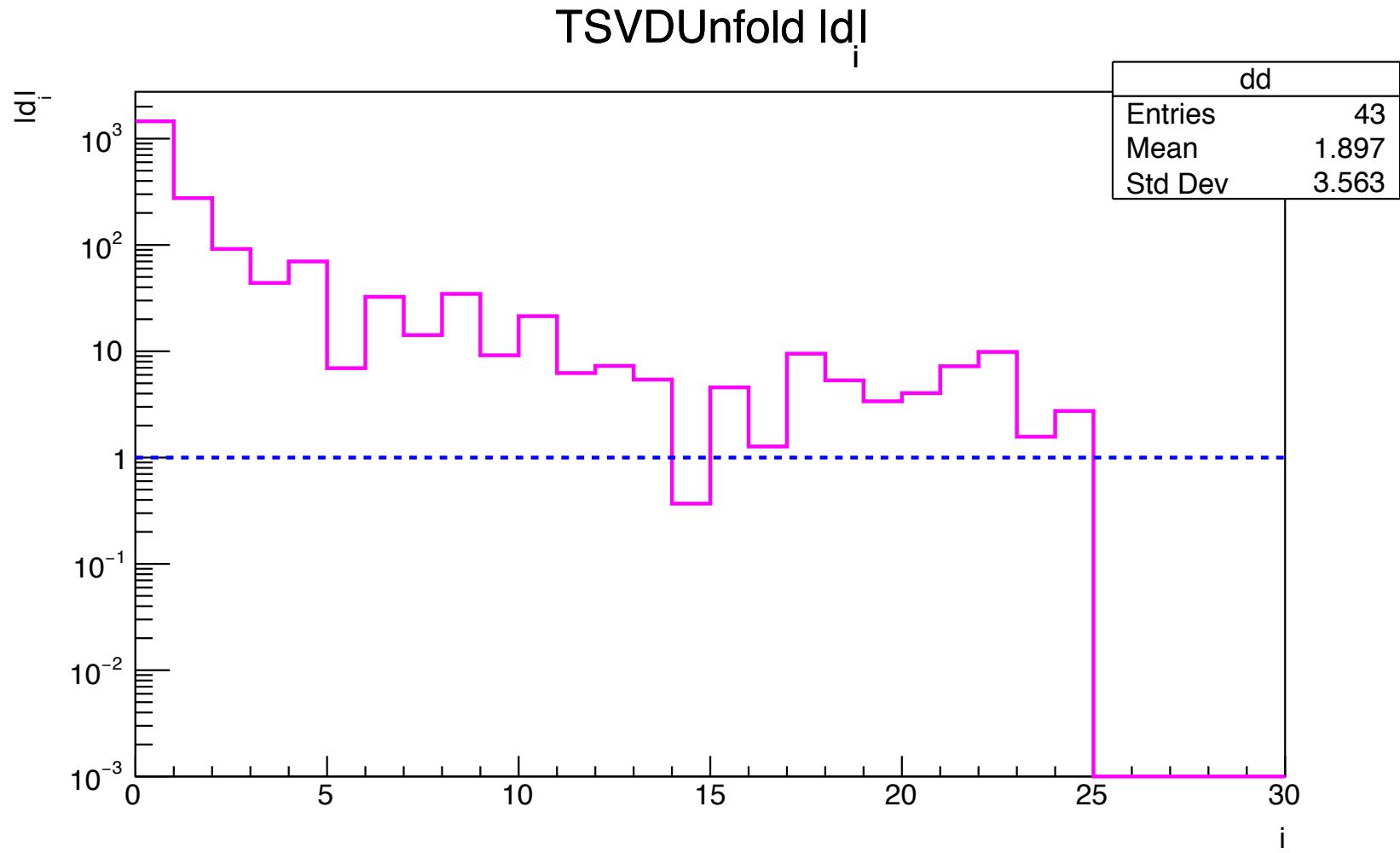
# Detector Response Matrix - UPC

Slide 13



# Detector Response Matrix – UPC d Vector

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# Unfolding Output – Unfolded Distribution

Slide 15

