

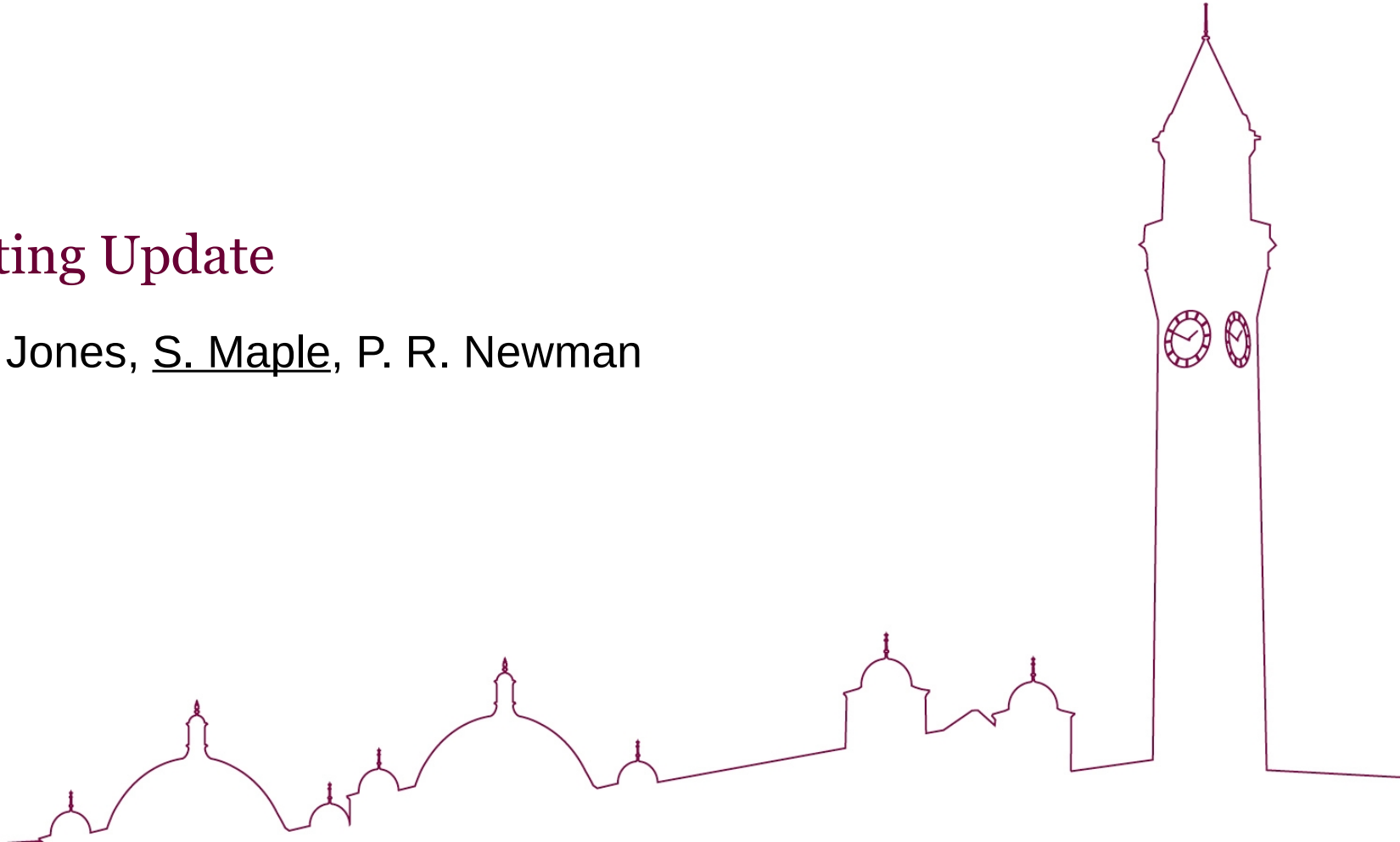


UNIVERSITY OF  
BIRMINGHAM

SCHOOL OF  
PHYSICS AND  
ASTRONOMY

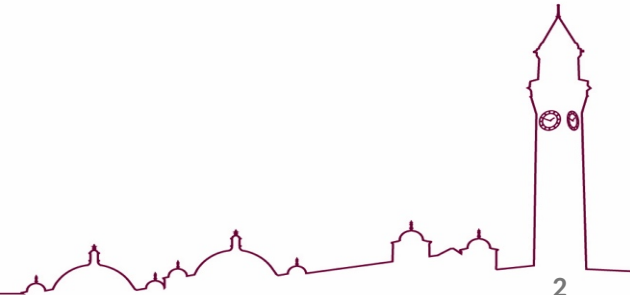
# Kinematic Fitting Update

L. Gonella, P. G. Jones, S. Maple, P. R. Newman



# Outline

- Performance as last shown
- Asymmetric Gaussian parametrisation of HFS
- More + Longer Markov Chains
- Extending to lower  $Q^2$



# Kinematic Fitting for DIS

- Only **need** 2 quantities to obtain  $\mathbf{x}$ ,  $\mathbf{y}$ ,  $Q^2$
- Using measured quantities  $\vec{\mathbf{D}} = \{\mathbf{E}_e, \theta_e, \delta_h, \mathbf{p}_{t,h}\}$  a kinematic fit can extract additional information:  $\vec{\lambda} = \{\mathbf{x}, \mathbf{y}, \mathbf{E}_\gamma\}$ 

$\mathbf{E}_\gamma$  is energy of an ISR photon
- For kinematic fit, can use a **likelihood** function based on knowledge of the detector resolutions:

## Likelihood

$$P(\vec{\mathbf{D}} | \vec{\lambda}) \propto \frac{1}{\sqrt{2\pi}\sigma_E} e^{-\frac{(E_e - E_e^\lambda)^2}{2\sigma_E^2}} \frac{1}{\sqrt{2\pi}\sigma_\theta} e^{-\frac{(\theta_e - \theta_e^\lambda)^2}{2\sigma_\theta^2}} \frac{1}{\sqrt{2\pi}\sigma_{\delta_h}} e^{-\frac{(\delta_h - \delta_h^\lambda)^2}{2\sigma_{\delta_h}^2}} \frac{1}{\sqrt{2\pi}\sigma_{P_{T,h}}} e^{-\frac{(P_{T,h} - P_{T,h}^\lambda)^2}{2\sigma_{P_{T,h}}^2}}$$

- Note: above quantities taken to be uncorrelated → Correlations between  $E_e$ ,  $\theta_e$  and  $\delta_h$ ,  $\mathbf{p}_{t,h}$  will later need to be taken into account

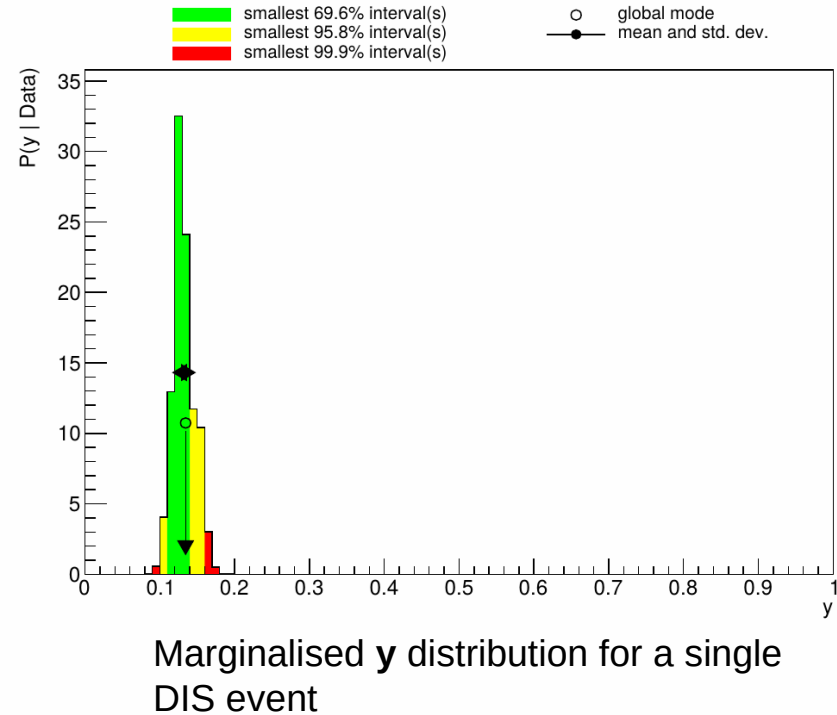
# Kinematic Fitting for DIS – A Bayesian Approach

- A Bayesian method can be applied in which basic features of the DIS cross section are encoded as a **prior**:

## Prior

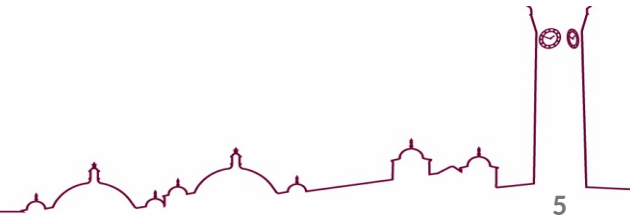
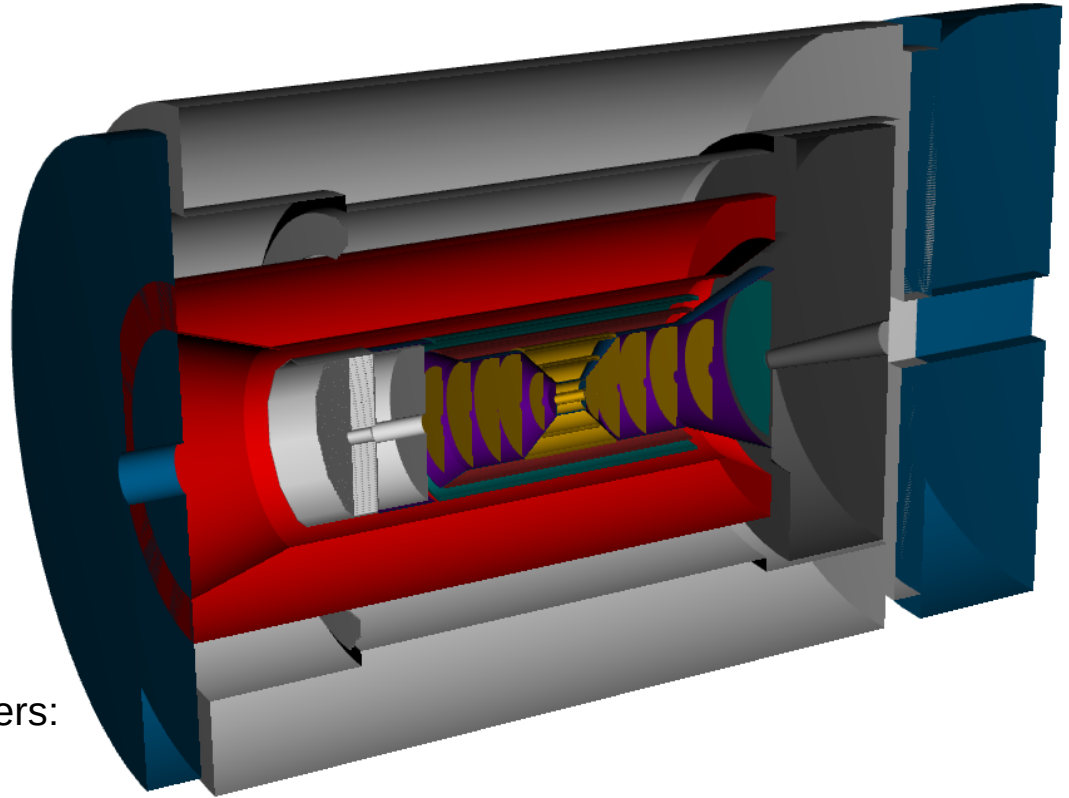
$$P_o(\vec{\lambda}) = \frac{1 + (1 - y)^2}{x^3 y^2} \frac{[1 + (1 - E_\gamma/A)^2]}{E_\gamma/A}$$

- Use “Bayesian analysis toolkit” to calculate most probable values of set  $\vec{\lambda}$  given measured quantities  $\vec{D}$ 
  - Metropolis algorithm MCMC used to find posterior distribution
  - Values for  $x$ ,  $y$ ,  $E_\gamma$  taken from global mode



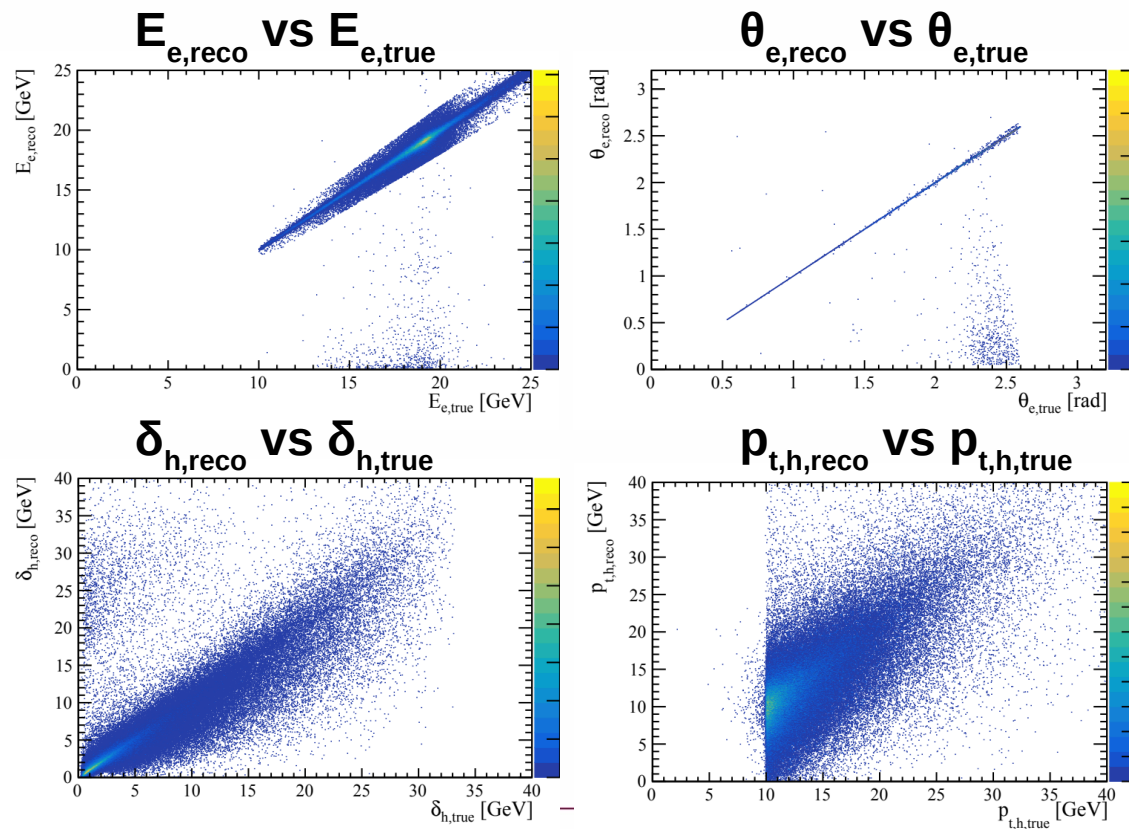
# ePIC Full Simulations

- Full simulation of Neutral Current DIS events in ePIC main detector
  - Pythia8 event generation
    - No QED corrections
    - $18 \times 275 \text{ GeV}^2$  e on p
    - $Q^2 > 100 \text{ GeV}^2$
    - Beam effects included
- Event reconstruction in Juggler
  - All charged particles taken from track measurements (ACTS)
  - Track projections matched to calorimeter clusters:
    - All clusters not associated with a track are added in as a neutral particle!



# Kinematic Fitting

- Choose events to be processed:
  - Select events where scattered electron is found in barrel with  $p_T > 10$  GeV (makes  $e^-$  finding easier)
  - Require  $y_{\text{true}} > 0.01$ : standard cut  $\rightarrow$  also ensures HFS well measured in hadron endcap



## Parametrisation

$$\sigma(E_e)/E_e = 0.02E_e \oplus 1.1\%$$

$$\sigma(\theta_e) = 1.1/E_e \oplus 0.05$$

$$\sigma(\delta_h) / \delta_h = 25\%$$

$$\sigma(p_{t,h}) / p_{t,h} = 25\%$$

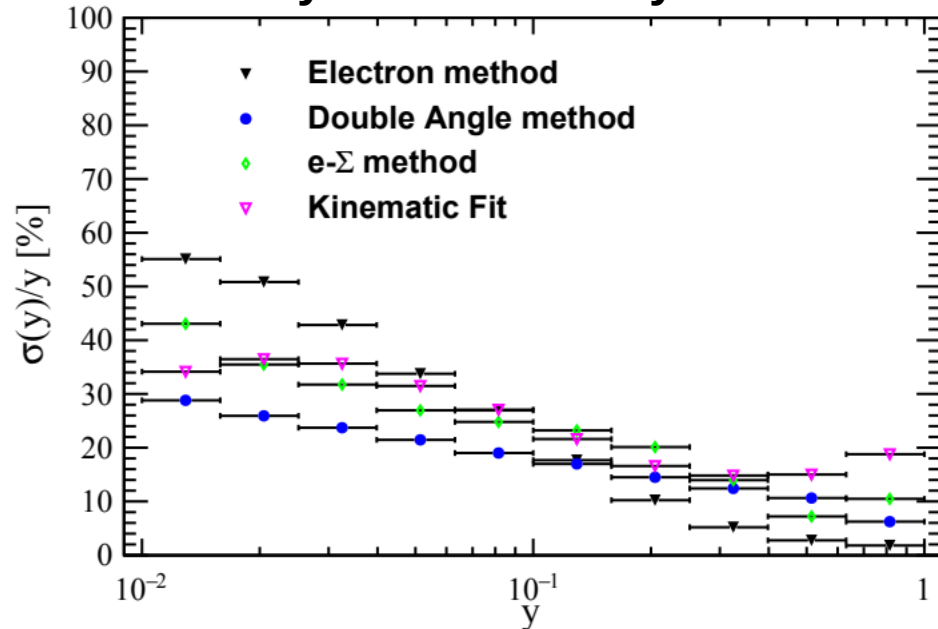
Note: Distributions from full simulation are complicated

$\rightarrow$  **Detailed studies needed**

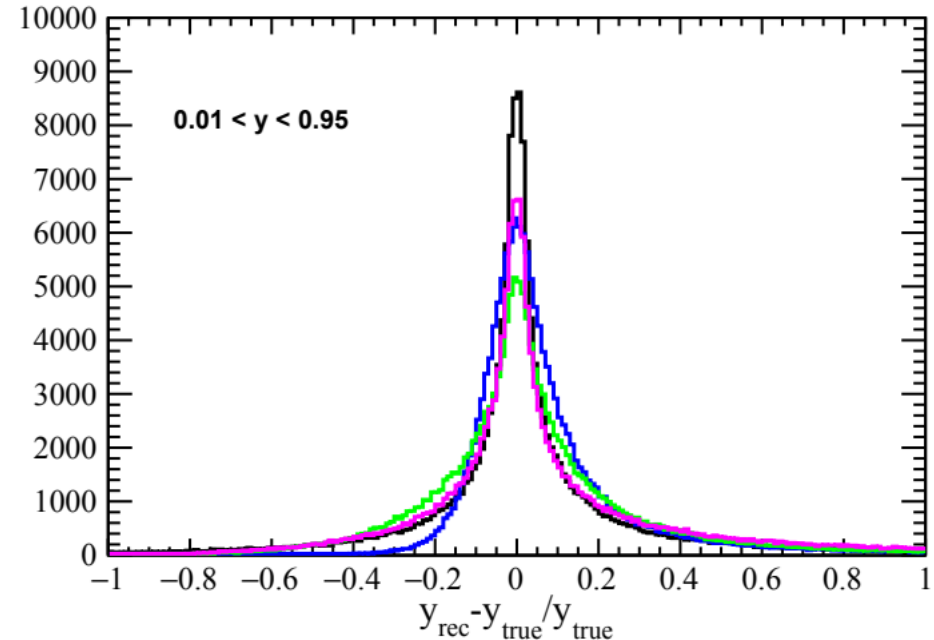
# Full Simulations: Fit Results

- Kinematic Fit of fully simulated data gives “okay” performance, at same level as conventional methods, but doesn’t win as expected
  - Possible causes: a) Inaccurate parametrisation of detector performance, or b) maximum likelihood value not being properly found... or both

**$y$  resolution vs  $y$**



**$y_{\text{rec}} - y_{\text{true}} / y_{\text{true}}$  for  $0.01 < y < 0.95$**



# Asymmetric HFS parametrisation

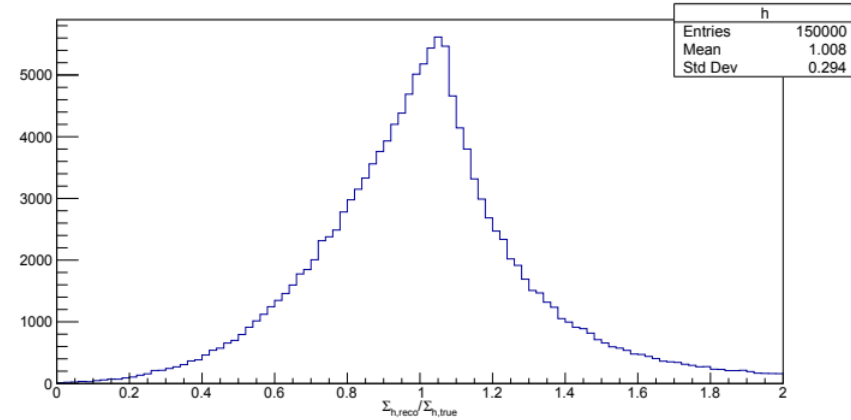
- Reconstructed/True distribution is clearly not a standard gaussian as assumed when calculating likelihood:

$$\frac{1}{\sqrt{2\pi}\sigma_{\delta_h}} e^{-\frac{(\delta_h - \delta_h^\lambda)^2}{2\sigma_{\delta_h}^2}} \frac{1}{\sqrt{2\pi}\sigma_{P_{T,h}}} e^{-\frac{(P_{T,h} - P_{T,h}^\lambda)^2}{2\sigma_{P_{T,h}}^2}}$$

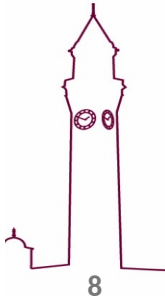
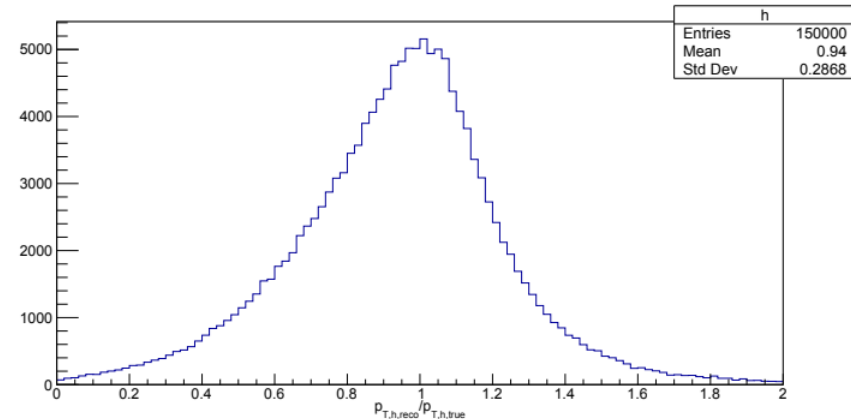
- Use split gaussian fit to calculate likelihood

$$\sigma(\delta_h) / \delta_h = 35\% \text{ (below) or } 25\% \text{ (above)}$$
$$\sigma(p_{t,h}) / p_{t,h} = 35\% \text{ (below) or } 25\% \text{ (above)}$$

- Measured E-p<sub>z</sub> divided by true



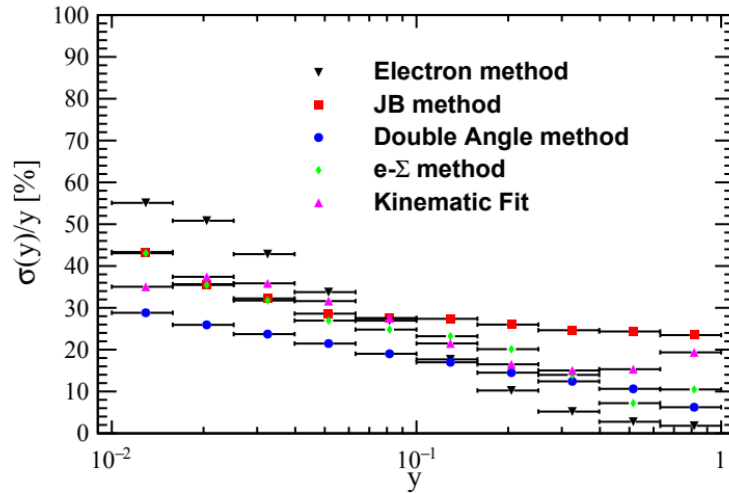
- P<sub>t</sub> predicted by fit divided by true





# Asymmetric HFS parametrisation

**RMS(y) vs y with split gaussian HFS parametrisation**

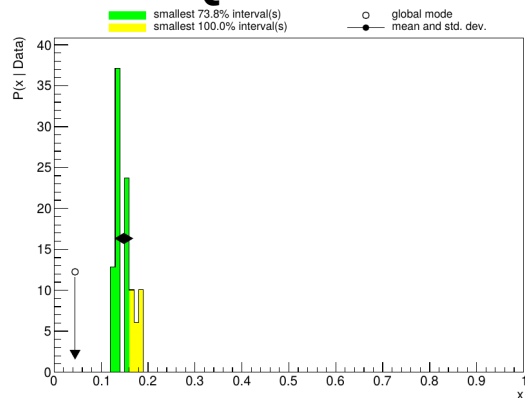


- Performance is essentially the same as before → pursue avenue b) (that the maximum likelihood value isn't being properly found)

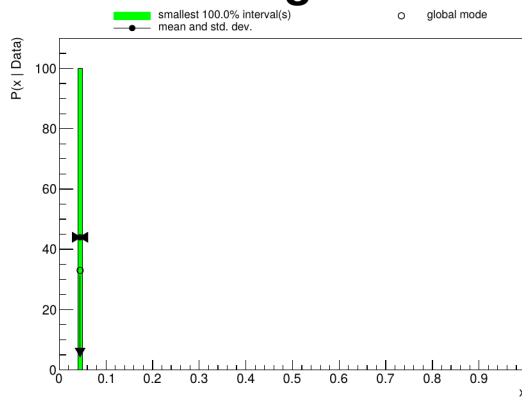
# (Marginalised) Posterior distributions vs precision

```
root [3] event var->Scan("x true:y true")
*****
*   Row   * x true *   y true *
*****
*         * 0.0441935 * 0.5189875 *
*****
```

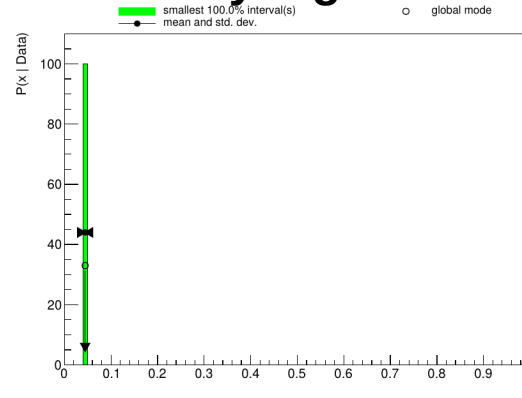
Quick



High

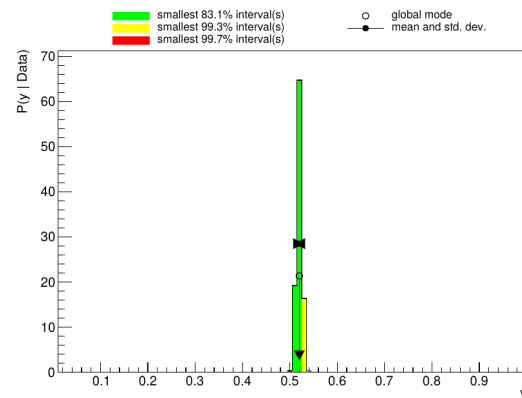
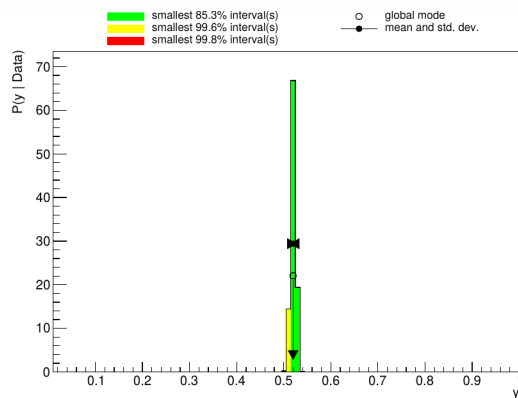
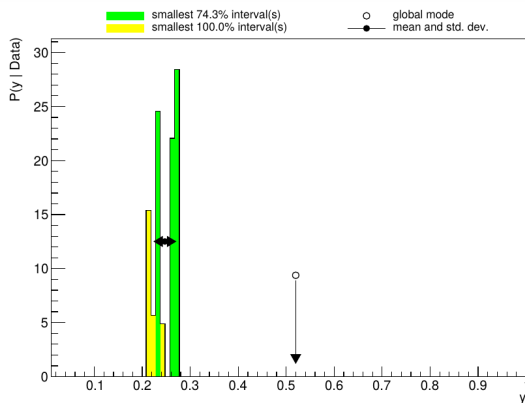


Very high



Previous results  
used Quick  
setting

High and Very  
High usually  
converge



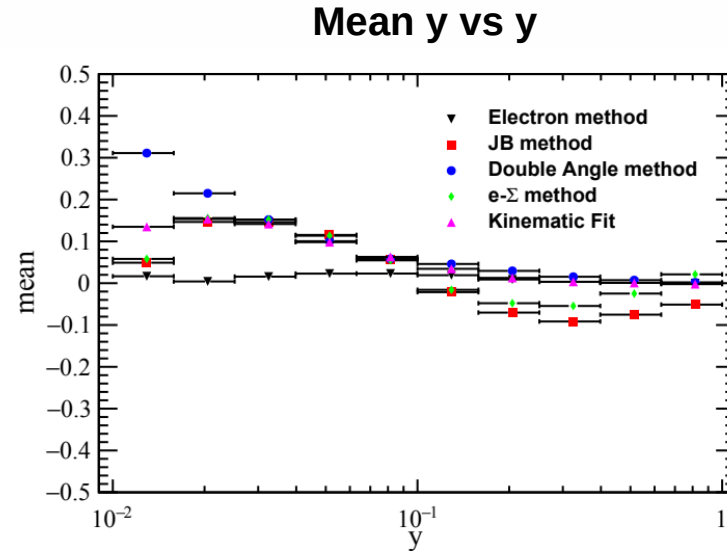
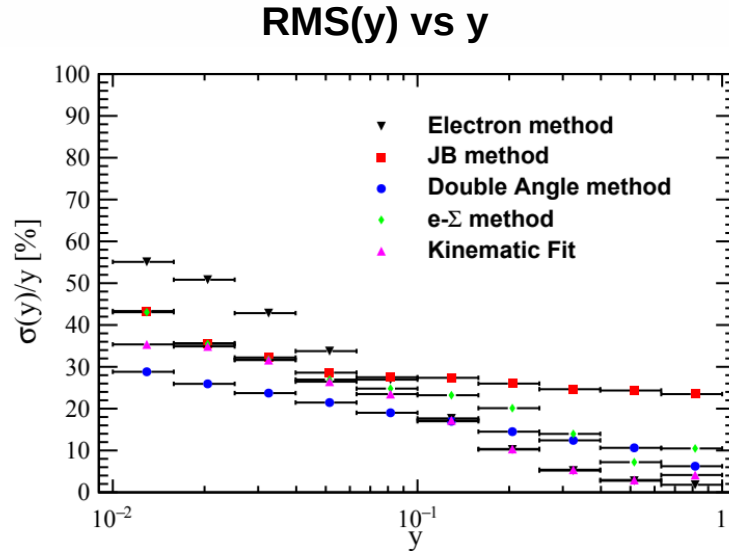
Very High  
setting runs  
slowly → High  
precision is a  
happy middle  
ground

2 Chains, 10k iterations

8 Chains, 1M iterations

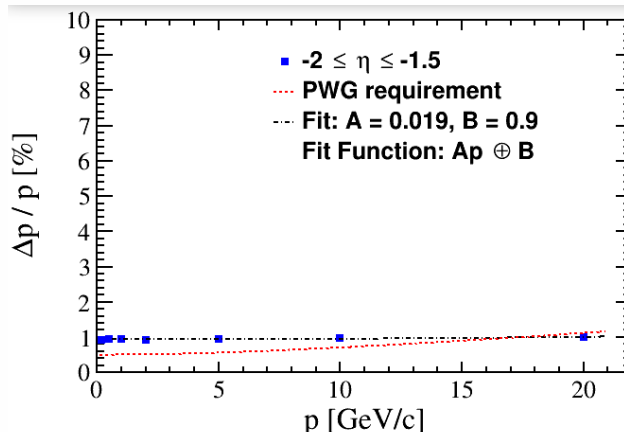
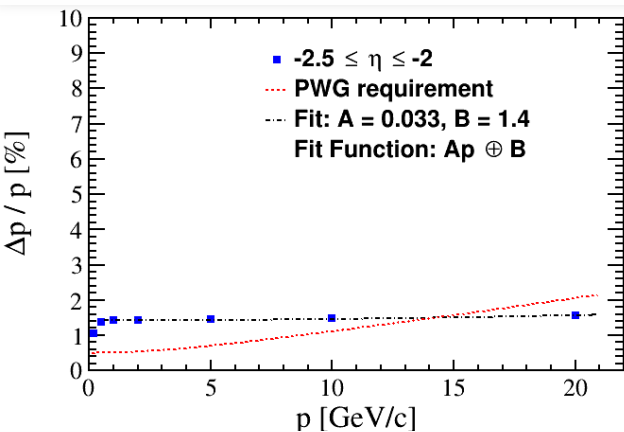
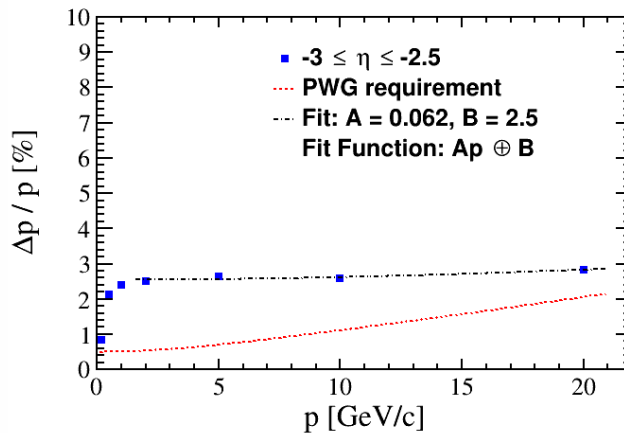
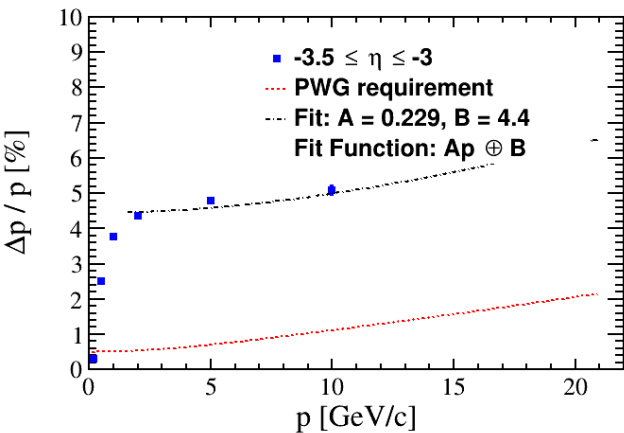
8 Chains, 10M iterations

# Fitting same events with “High precision”



- Significant improvement to KF resolution!
- Generally matches or beats resolution of methods other than DA, which has better resolution at lower y (but larger mean offset)

# Extending to lower $Q^2$

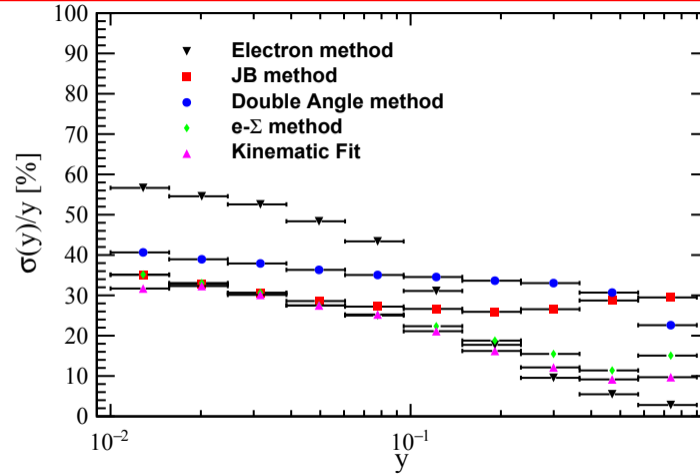
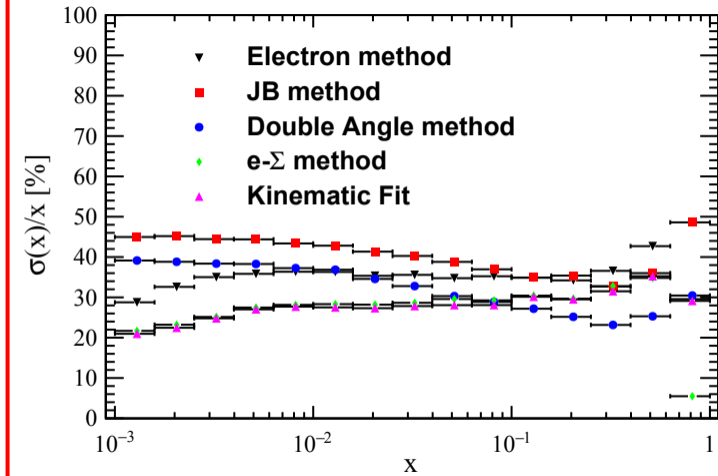


- Previously restricted events to high  $Q^2$  events with electrons scattered into barrel
- Extended to events with  $Q^2 > 1 \text{ GeV}^2$  → Requires parametrisation of  $dE/E$  and  $d\theta$  in pseudorapidity bins

## A couple of caveats:

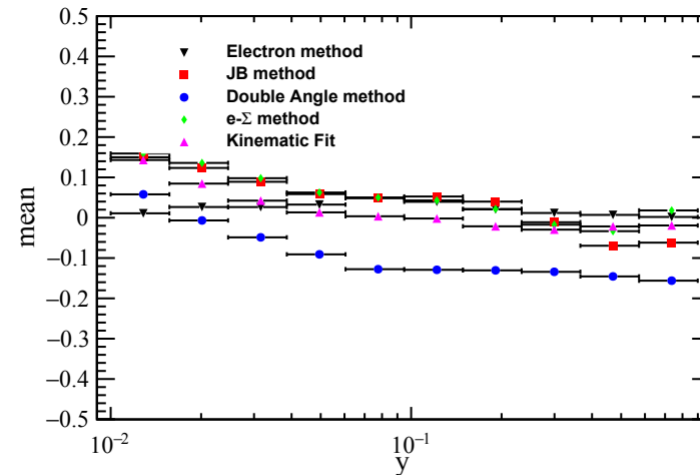
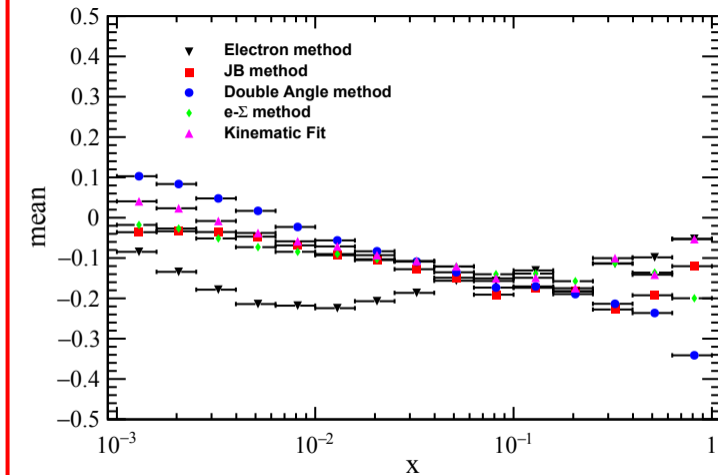
- At low  $p_T$  an issue with truth track seeding in simulations at the time results sees  $dp/p$  improve at low  $p$  → unphysical (“fixed” in eicrecon)
- Electron “finding” wasn’t changed in my recon → bad approximation at high  $y$

# Resolutions (Juggler $Q^2 > 1\text{GeV}^2$ )



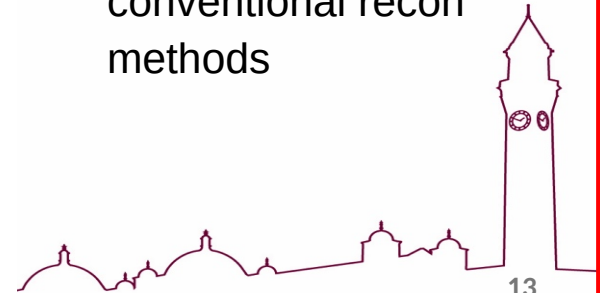
## Resolution

- KF matches or beats conventional recon methods except e-method at high  $y$  (poorly described due to previously mentioned problems)



## Mean

- KF shows low level of bias compared to conventional recon methods



# Future Plans

- EIC Simulation and Reconstruction software are being (rapidly) developed, but there is still much work to be done for complete realism
- I now have an H1 computing account (thanks H1 folks!) → plan to use this to test KF with mature MC, and at some point with real H1 data
- Thanks to Barak and Kolja we now have Djangoh files successfully passing through afterburner, which should be tested with full sim (I tried a while ago but simulation crashed for me) → can hopefully figure this out and have some full sim events with QED=on to test with KF to show full power of the method
- Should eventually look at correlations between measured quantities e.g.  $p_T$  and  $\Sigma_h$  and include these in KF

