

Modern jet physics with AI

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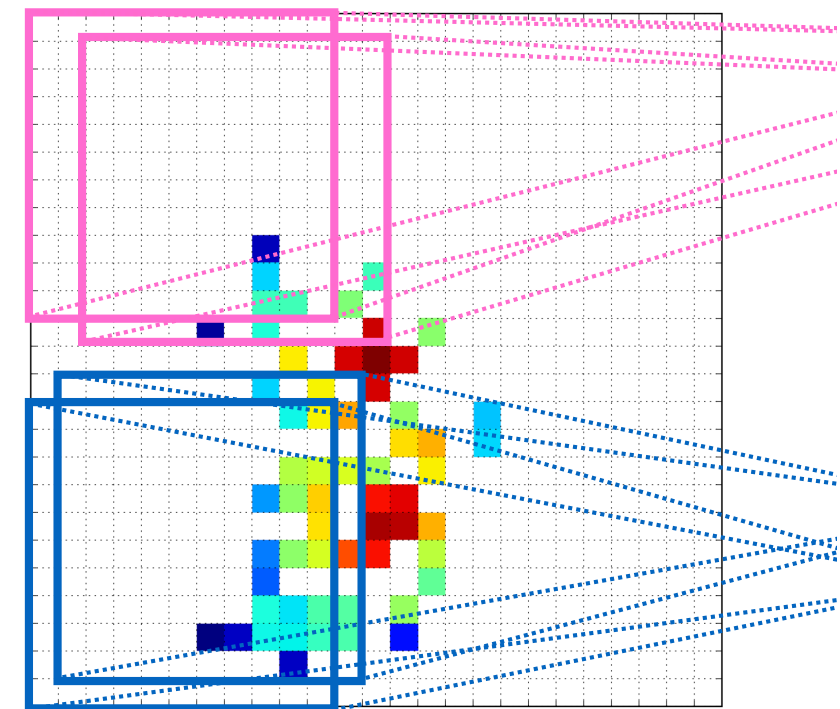
bpnachman@lbl.gov



@bpnachman



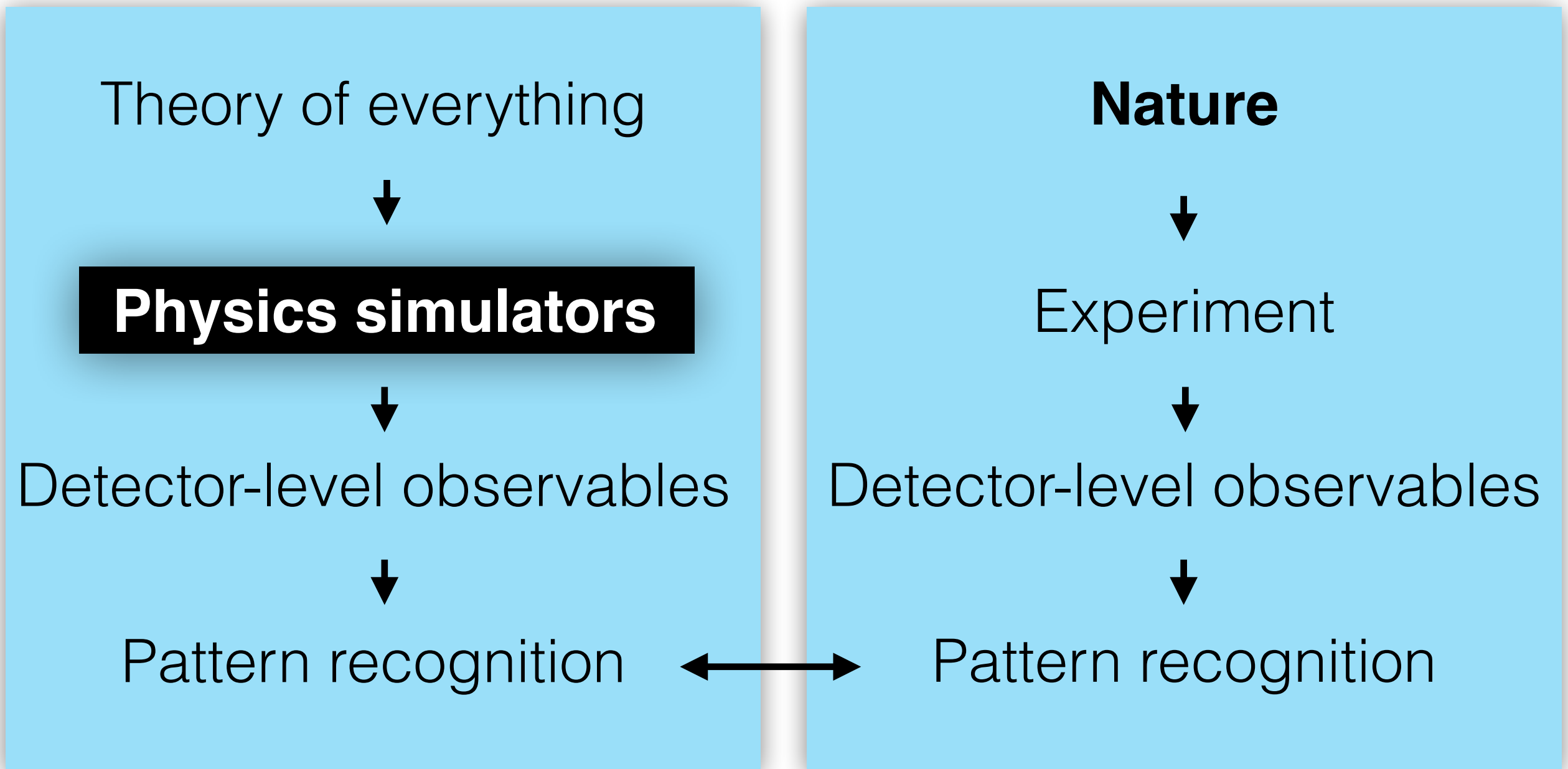
bnachman



Future *ep* workshop
Oct. 26, 2022

Data analysis in High Energy Physics

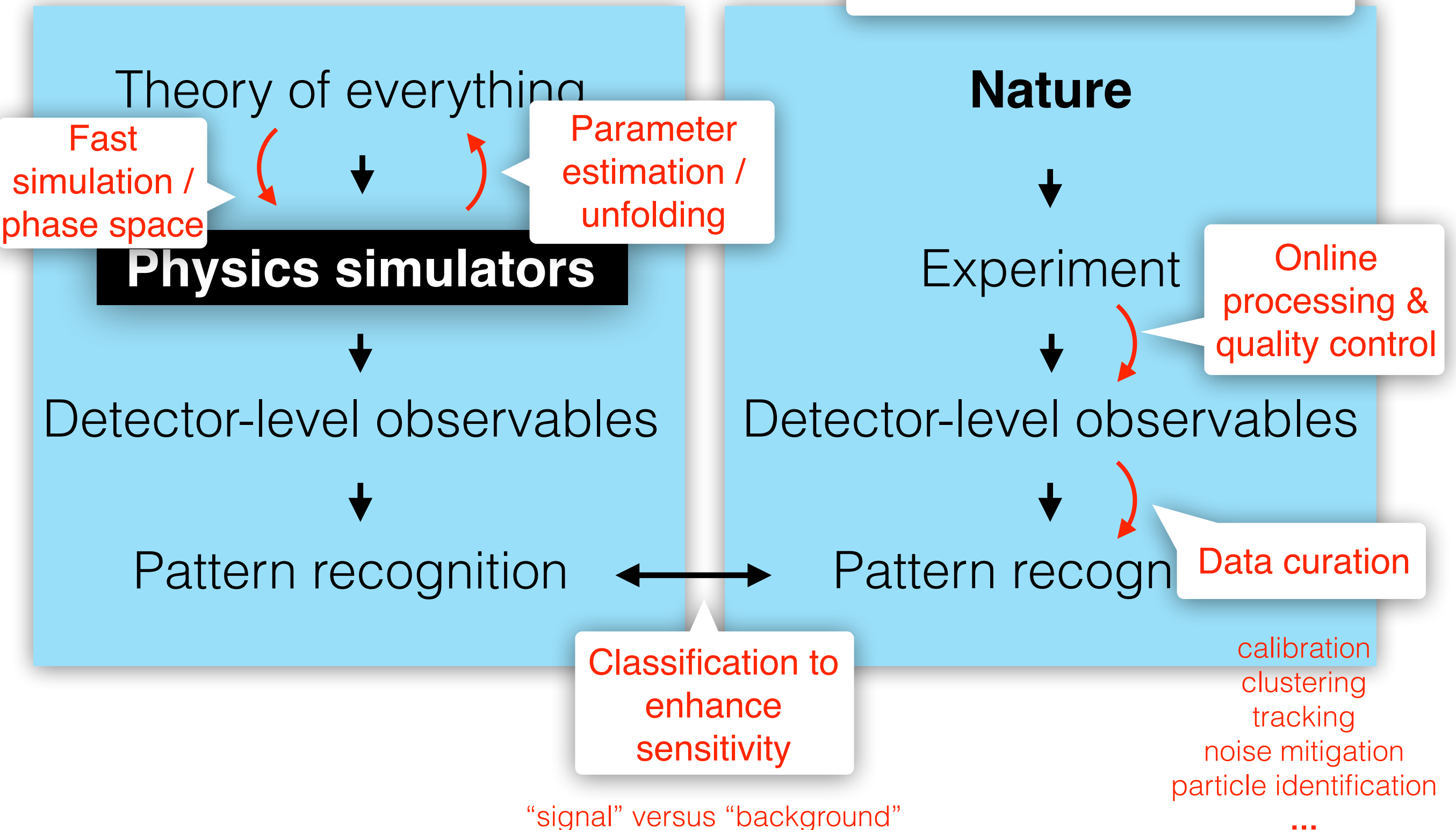
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Data analysis in High Energy Physics

3

+ Machine Learning



Data analysis in High Energy Physics

4

+ Machine Learning

Today: I'll only have time to cover two topics:

Reconstruction and **Unfolding**

Fast
simulation /
phase space

Parameter
estimation /
unfolding

Physics simulators

Detector-level observables

Pattern recognition



enhance



"signal" versus "background"

Experiment

Online
processing &
quality control

Detector-level observables

Pattern recognition

Data curation

calibration
clustering
tracking
noise mitigation
particle identification
...

Part I: Reconstruction



Reconstruction



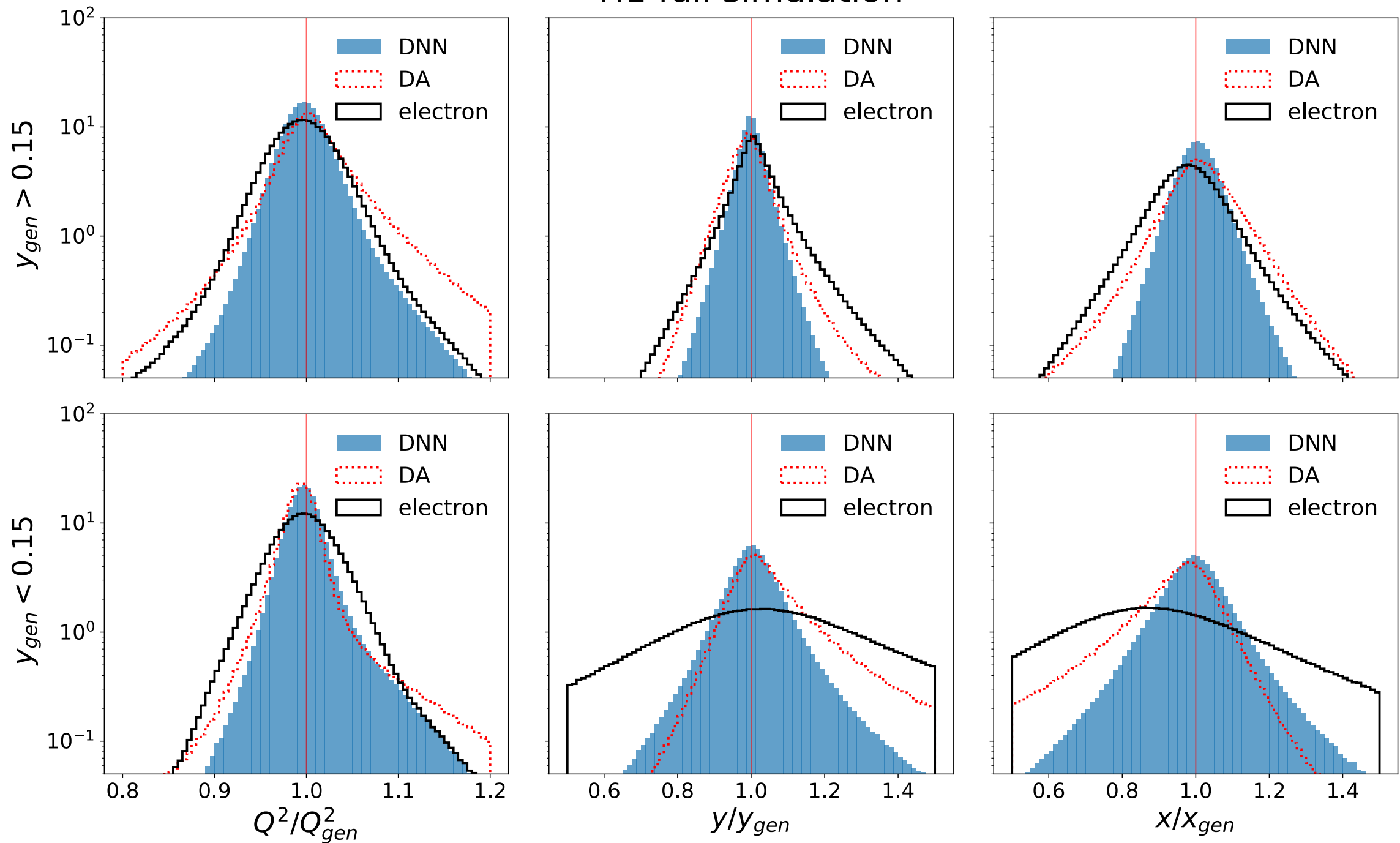
Inclusive DIS; not jets per se, but critical for all jet studies

Method name	Observables	y	Q^2	$x \cdot E_p$
Electron (e)	$[E_0, E, \theta]$	$1 - \frac{\Sigma_e}{2E_0}$	$\frac{E^2 \sin^2 \theta}{1-y}$	$\frac{E(1+\cos \theta)}{2y}$
Double angle (DA) [6, 7]	$[E_0, \theta, \gamma]$	$\frac{\tan \frac{\gamma}{2}}{\tan \frac{\gamma}{2} + \tan \frac{\theta}{2}}$	$4E_0^2 \cot^2 \frac{\theta}{2} (1-y)$	$\frac{Q^2}{4E_0 y}$
Hadron (h , JB) [4]	$[E_0, \Sigma, \gamma]$	$\frac{\Sigma}{2E_0}$	$\frac{T^2}{1-y}$	$\frac{Q^2}{2\Sigma}$
ISigma ($I\Sigma$) [9]	$[E, \theta, \Sigma]$	$\frac{\Sigma}{\Sigma + \Sigma_e}$	$\frac{E^2 \sin^2 \theta}{1-y}$	$\frac{E(1+\cos \theta)}{2y}$
IDA [7]	$[E, \theta, \gamma]$	y_{DA}	$\frac{E^2 \sin^2 \theta}{1-y}$	$\frac{E(1+\cos \theta)}{2y}$
$E_0 E \Sigma$	$[E_0, E, \Sigma]$	y_h	$4E_0 E - 4E_0^2 (1-y)$	$\frac{Q^2}{2\Sigma}$
$E_0 \theta \Sigma$	$[E_0, \theta, \Sigma]$	y_h	$4E_0^2 \cot^2 \frac{\theta}{2} (1-y)$	$\frac{Q^2}{2\Sigma}$
$\theta \Sigma \gamma$ [8]	$[\theta, \Sigma, \gamma]$	y_{DA}	$\frac{T^2}{1-y}$	$\frac{Q^2}{2\Sigma}$
Double energy (A4) [7]	$[E_0, E, E_h]$	$\frac{E-E_0}{(xE_p)-E_0}$	$4E_0 y (xE_p)$	$E + E_h - E_0$
$E \Sigma T$	$[E, \Sigma, T]$	$\frac{\Sigma}{\Sigma + E \pm \sqrt{E^2 + T^2}}$	$\frac{T^2}{1-y}$	$\frac{Q^2}{2\Sigma}$
$E_0 E T$	$[E_0, E, T]$	$\frac{2E_0 - E \mp \sqrt{E^2 - T^2}}{2E_0}$	$\frac{T^2}{1-y}$	$\frac{Q^2}{4E_0 y}$
Sigma (Σ) [9]	$[E_0, E, \Sigma, \theta]$	$y_{I\Sigma}$	$Q_{I\Sigma}^2$	$\frac{Q^2}{4E_0 y}$
eSigma ($e\Sigma$) [9]	$[E_0, E, \Sigma, \theta]$	$\frac{2E_0 \Sigma}{(\Sigma + \Sigma_e)^2}$	$2E_0 E (1 + \cos \theta)$	$\frac{E(1+\cos \theta)(\Sigma + \Sigma_e)}{2\Sigma}$

DIS Reco with simple NN



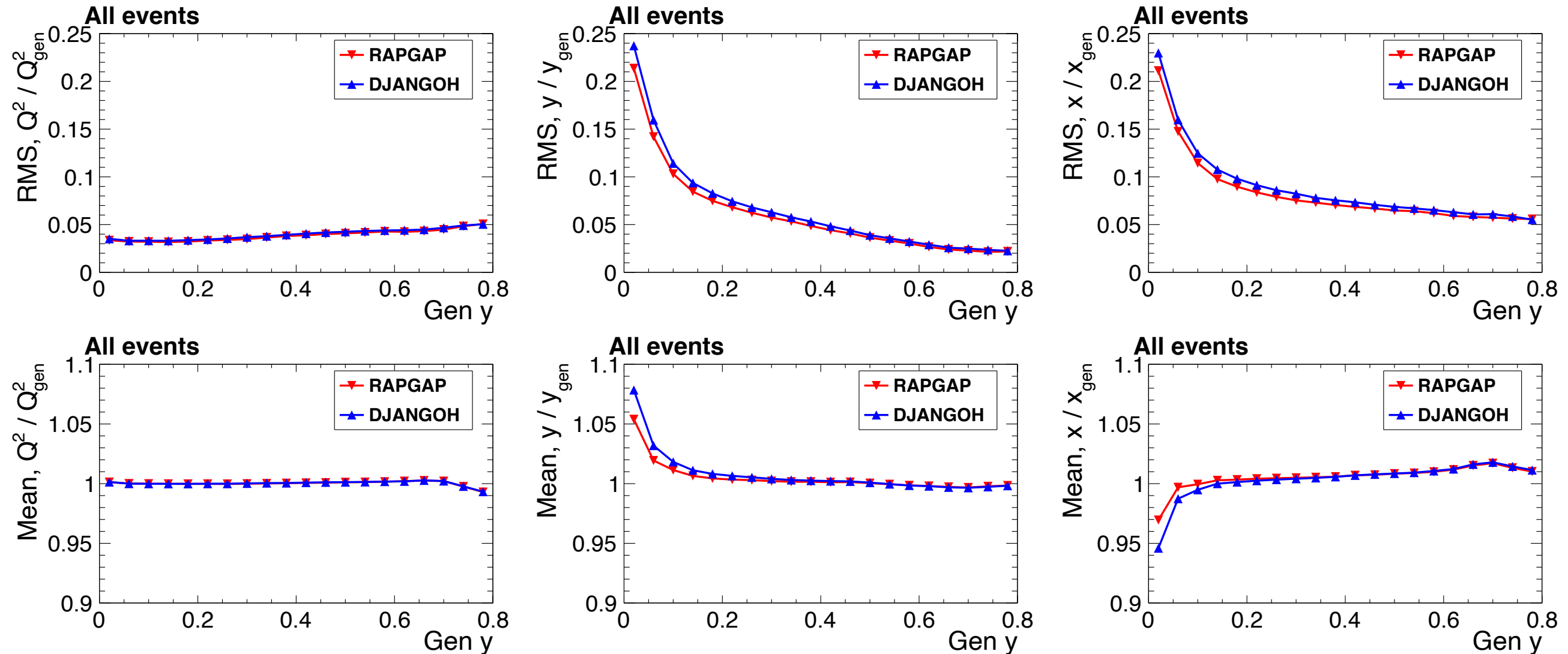
H1 full simulation



DIS Reco with simple NN



H1 full simulation



Holds up to model variations!

Reconstruction - word of caution!



All of the methods studied so far for DIS are of the form: predict true from measured via mean-squared error (or similar)

Claim: this is prior dependent !

What goes wrong?

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Suppose you have some features x and you want to predict y .

detector energy

true energy

One way to do this is to find an f that minimizes the mean squared error (MSE):

$$f = \operatorname{argmin}_g \sum_i (g(x_i) - y_i)^2$$

Then, $f(x) = E[y|x]$.

Why is this a problem?

What goes wrong?

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Suppose you have some features x and you want to predict y .

detector energy

true energy

$$f(x) = E[y|x] = \int dy y p(y|x)$$

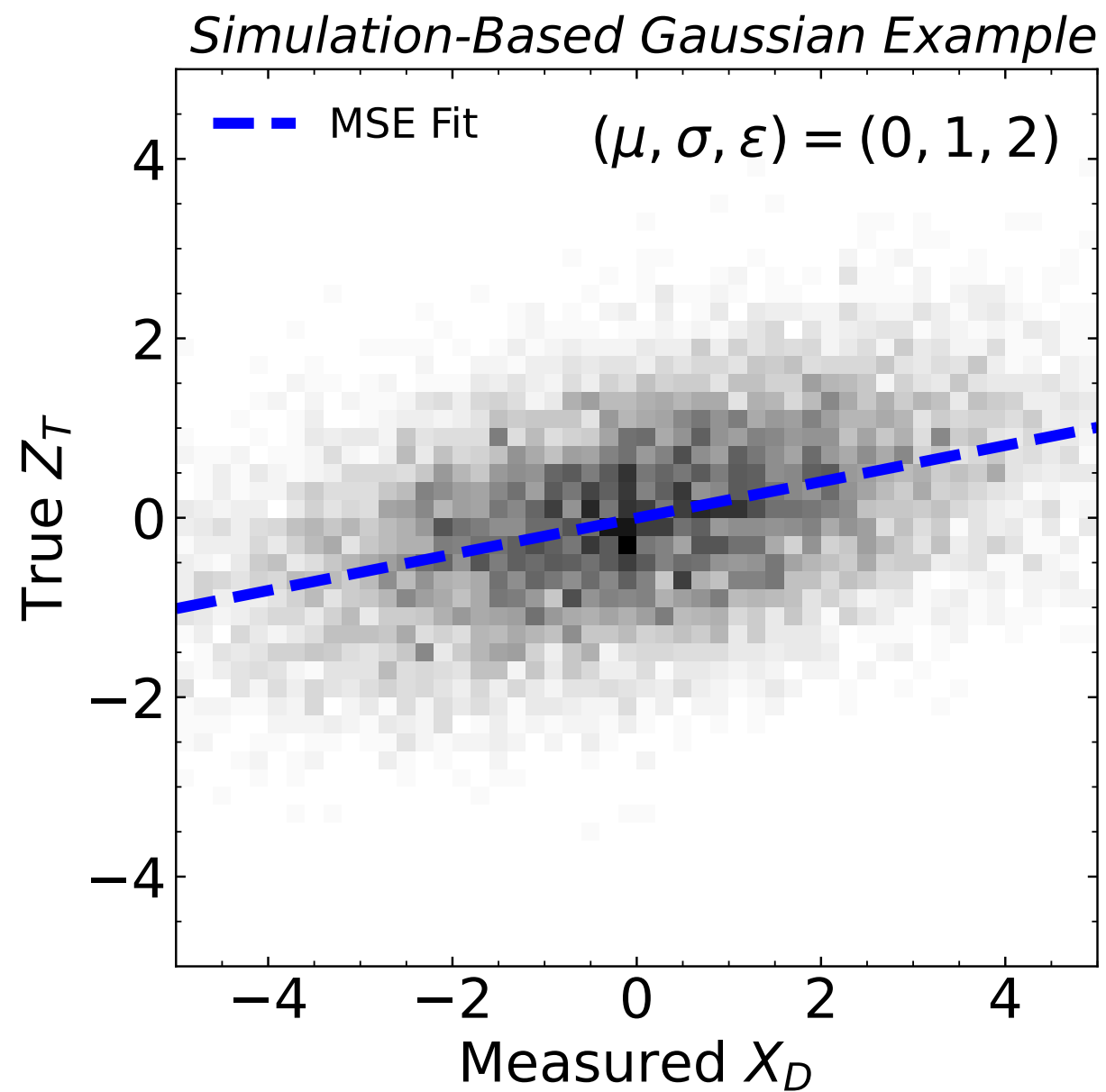
$$E[f(x)|y] = \int dx dy' y' p_{\text{train}}(y'|x) p_{\text{test}}(x|y)$$

this need not be y even if $p_{\text{train}} = p_{\text{test}}$ (!)

Why is this a problem?

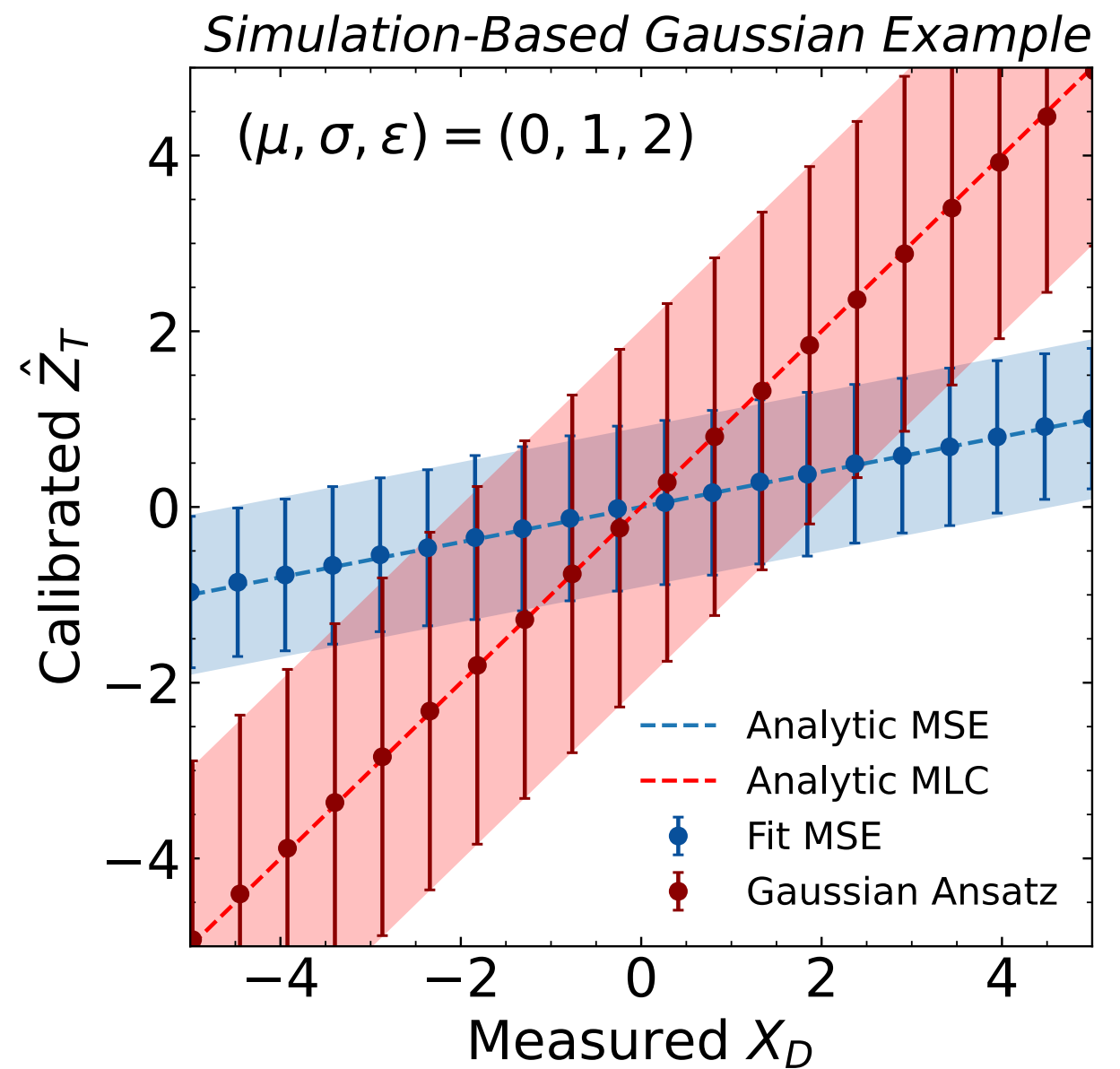
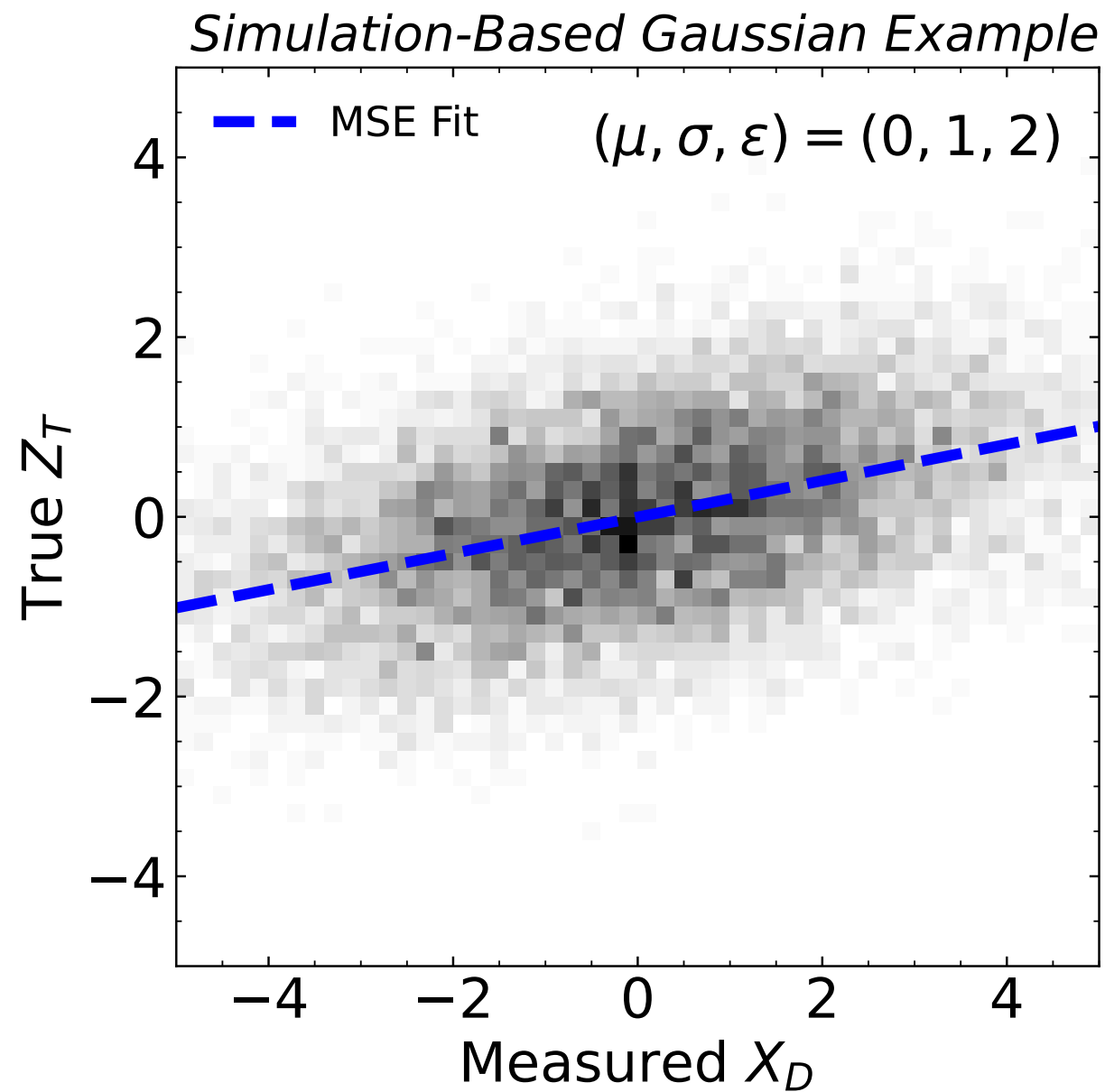
Gaussian Example

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Gaussian Example

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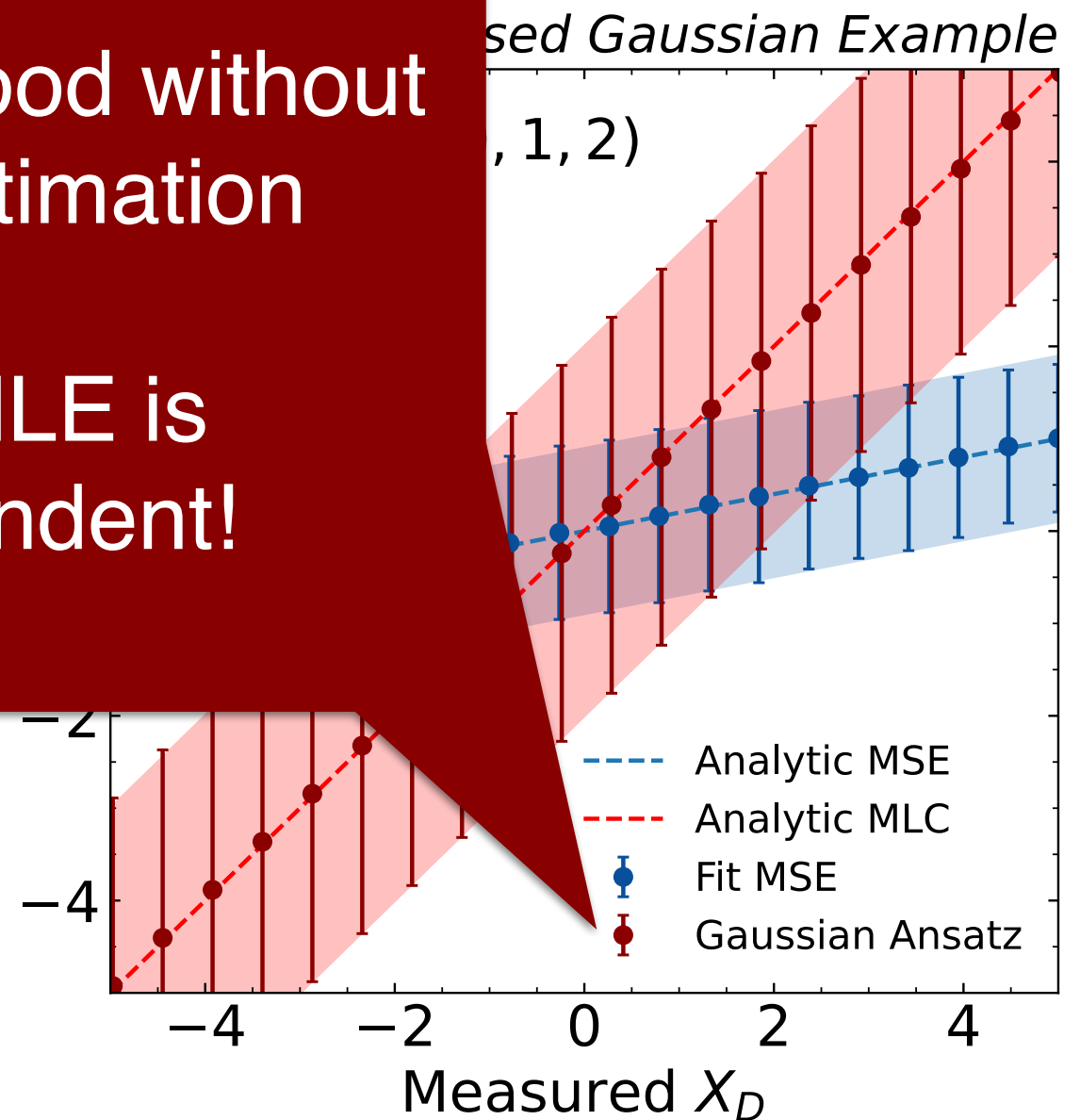
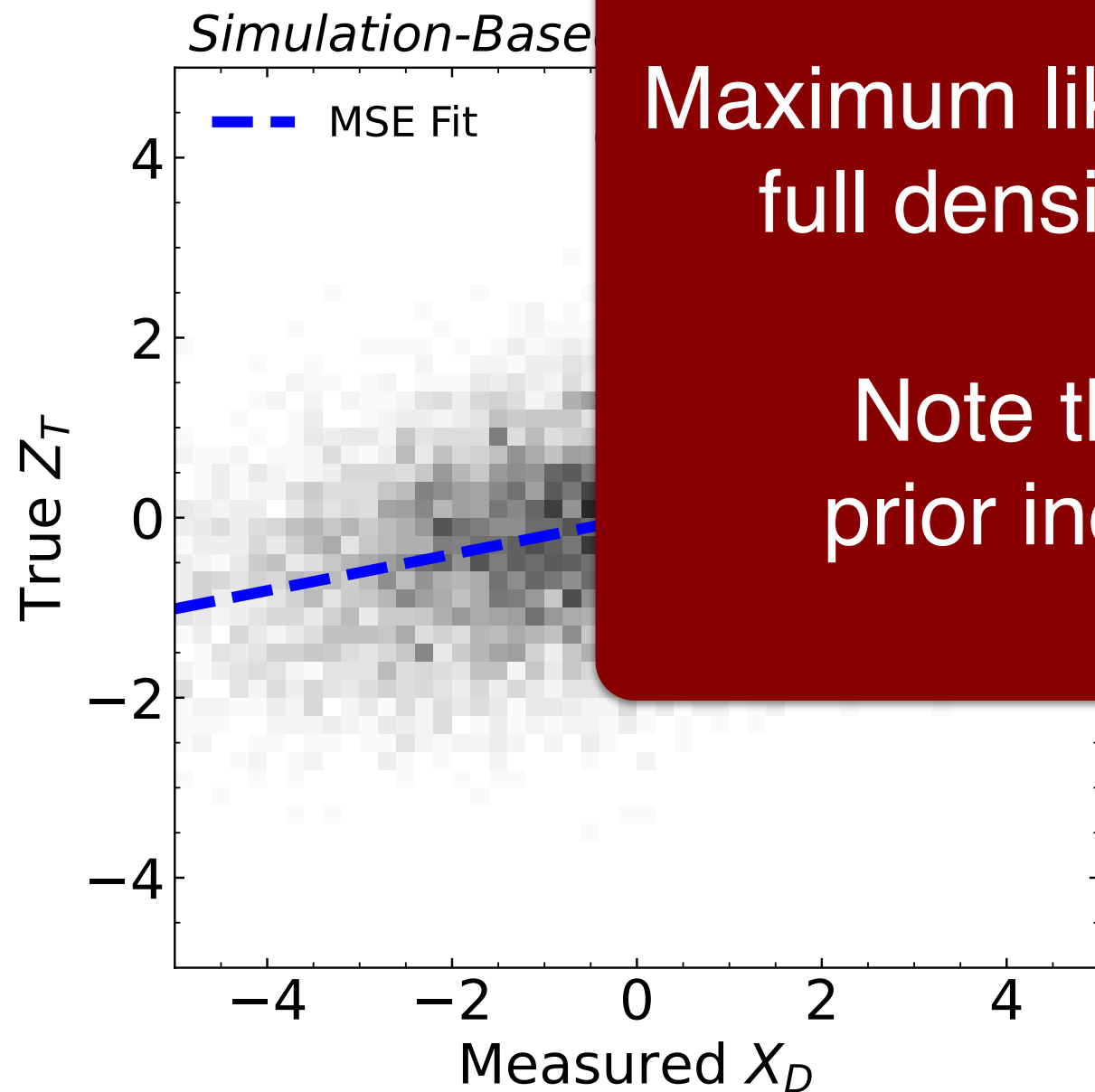


Gaussian Example

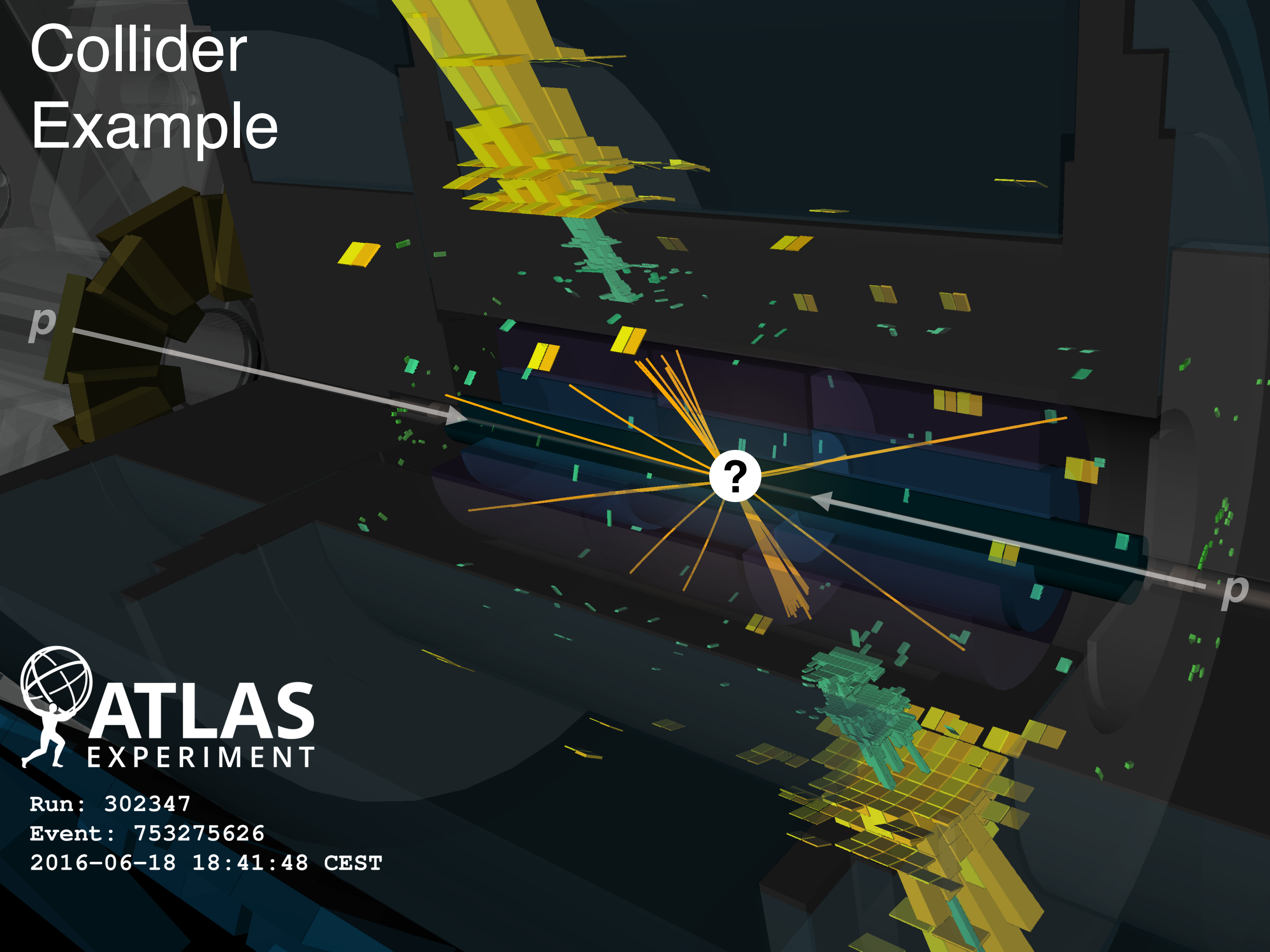
14

Maximum likelihood without
full density estimation

Note that MLE is
prior independent!



Collider Example



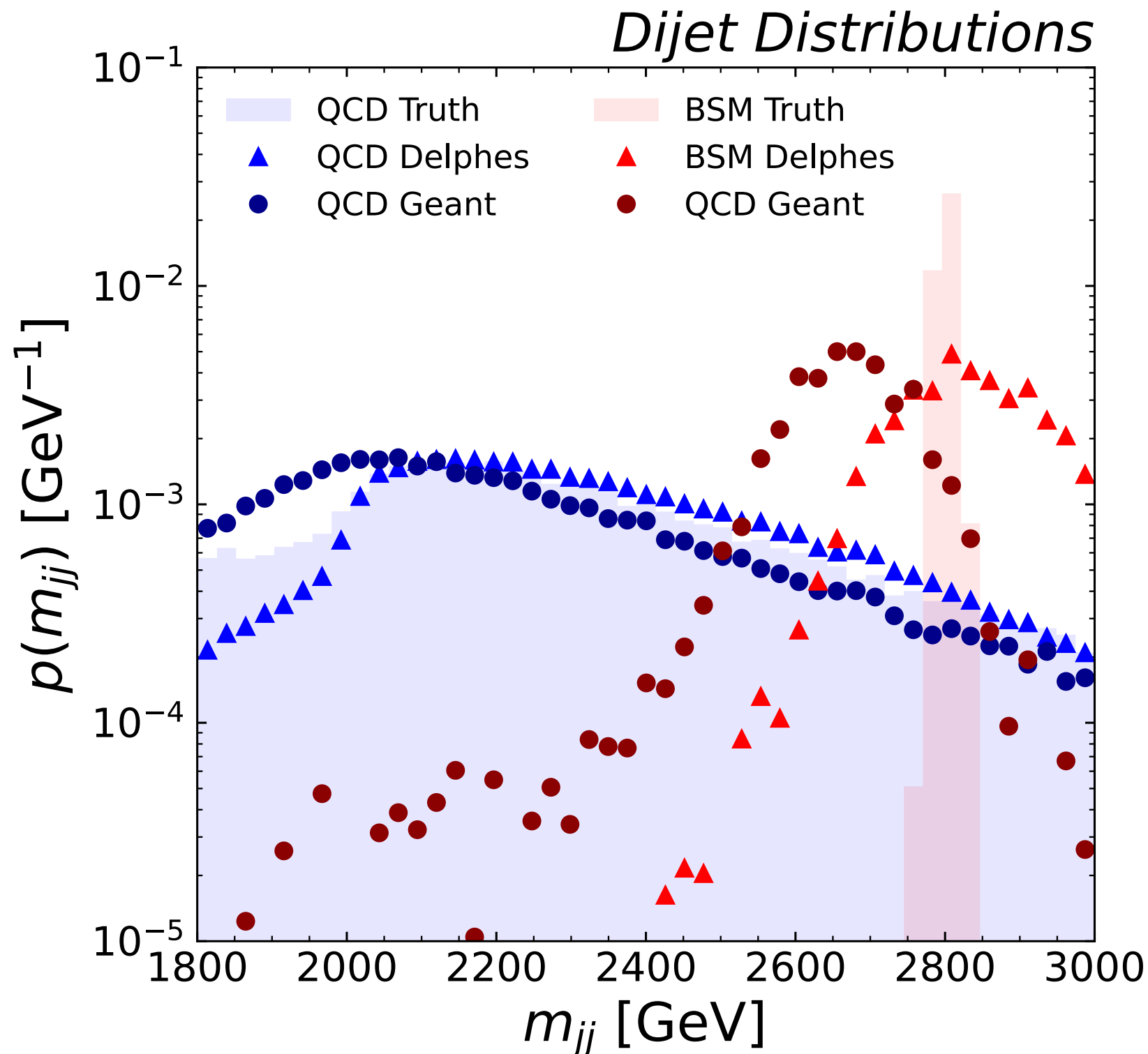
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Event: 753275626

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Collider Example

16



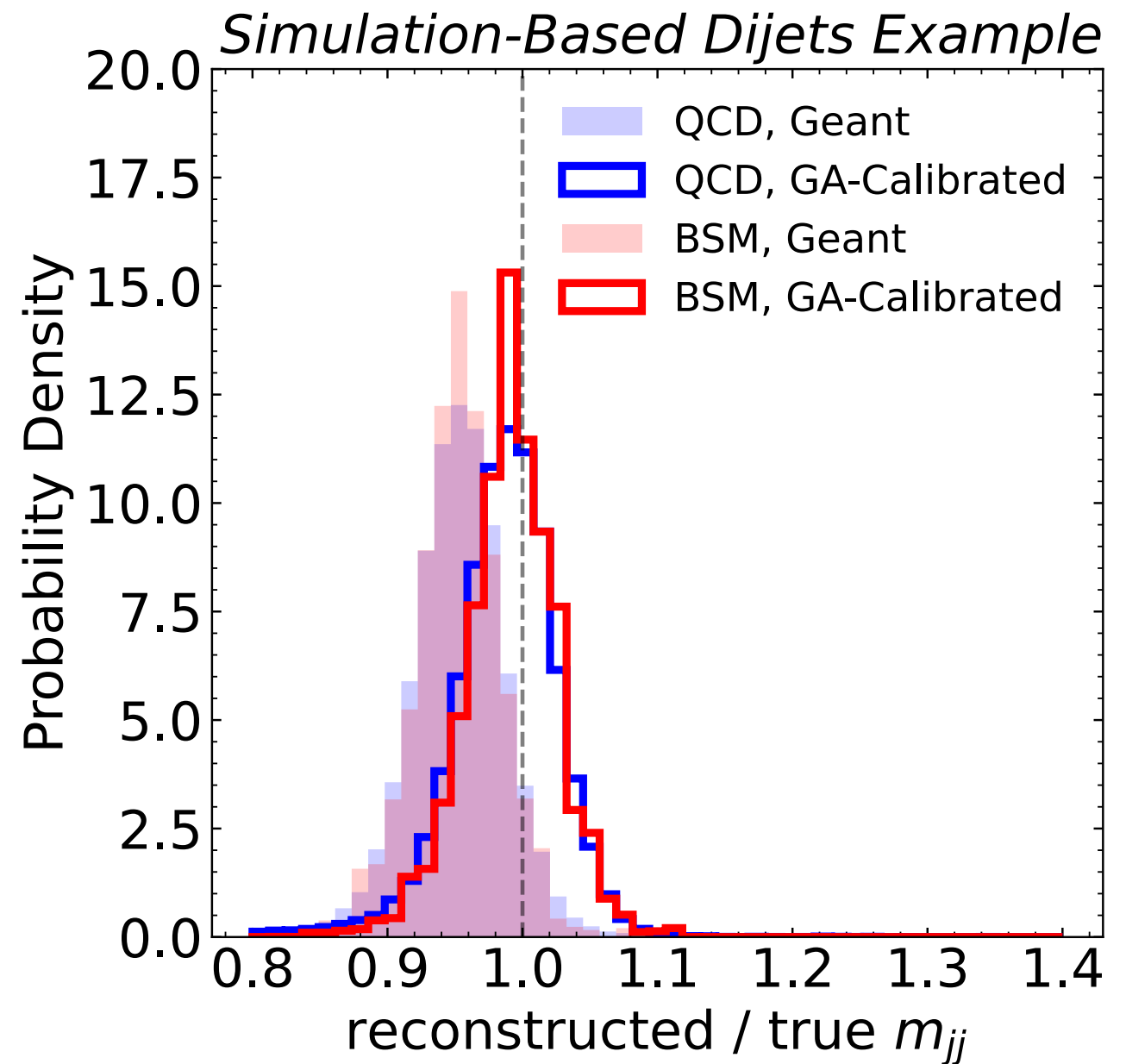
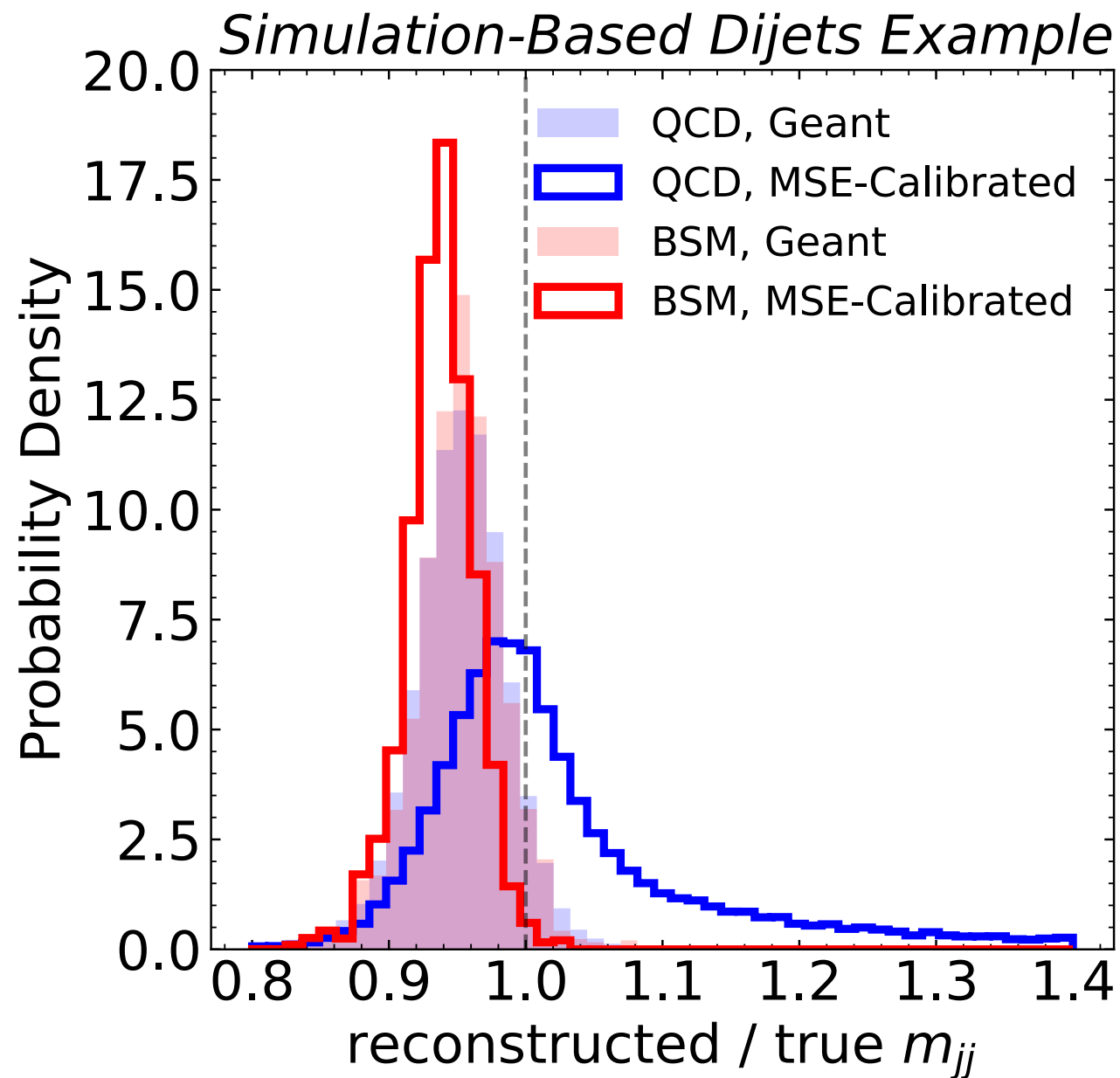
QCD = quarks
and gluons

BSM = new
physics

Looking for new
massive particles that
produce jets

Collider Example

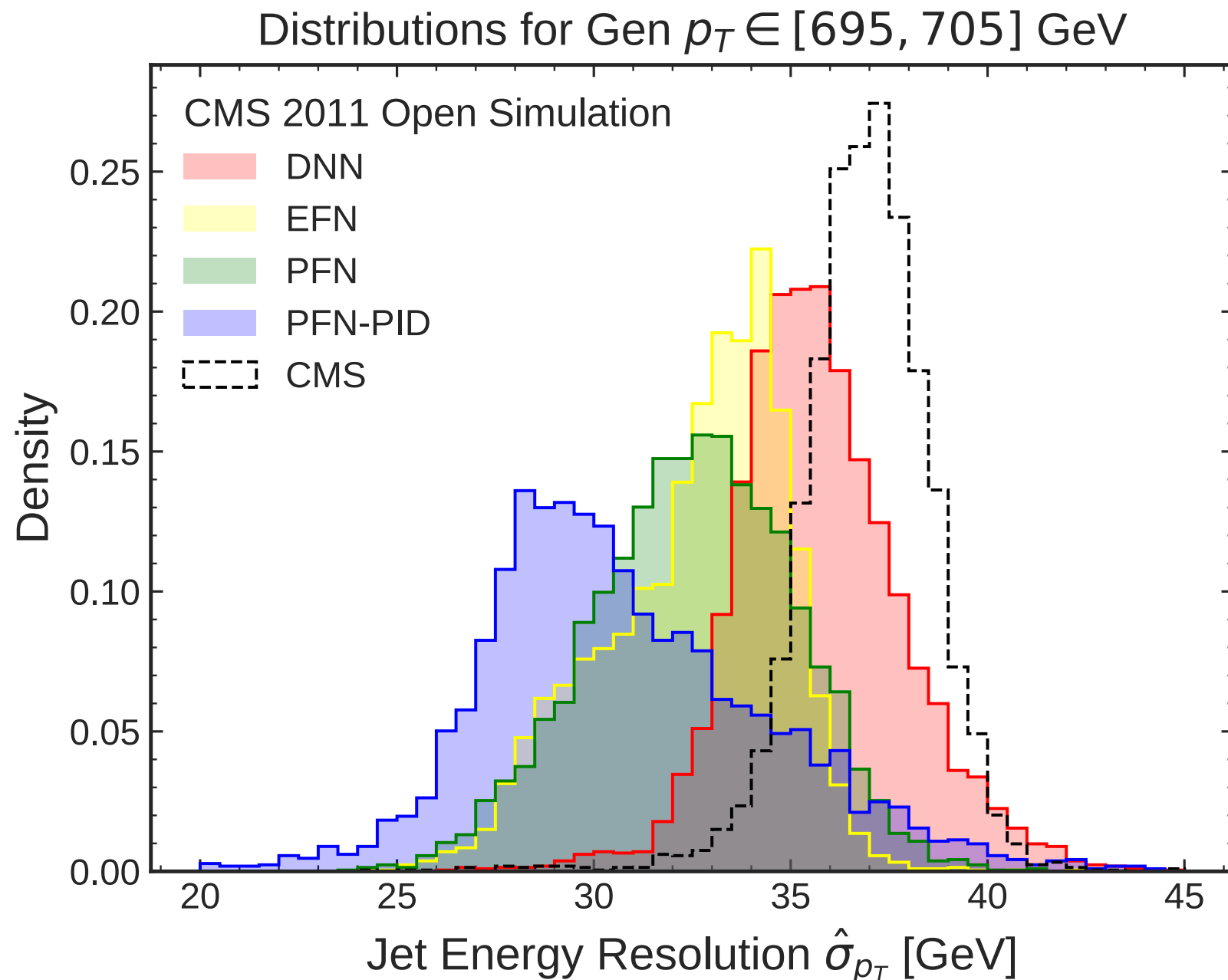
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Collider Example

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Can do high-dimensional regression in this framework



PFN-ID: process all particles inside jets

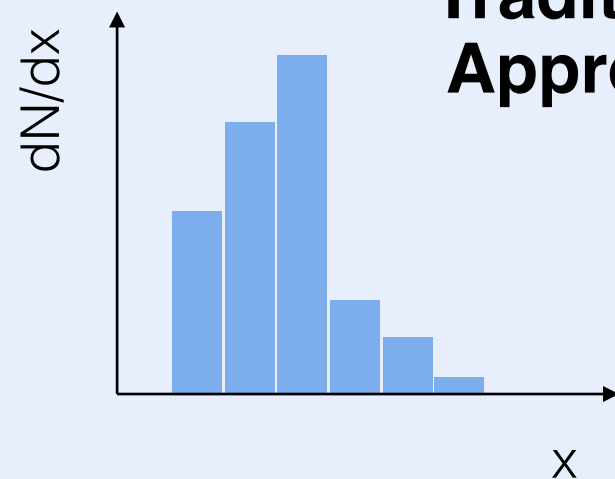
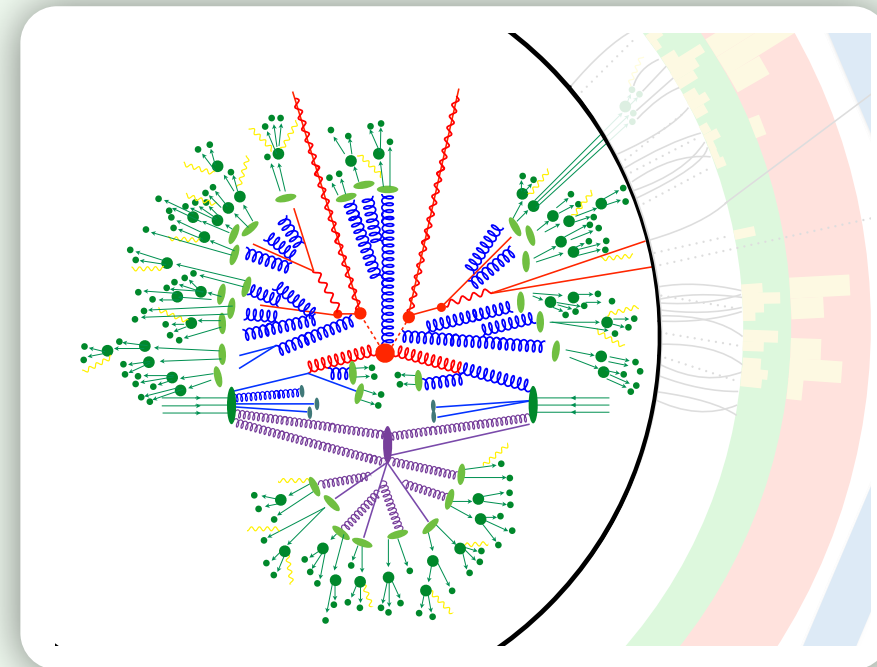
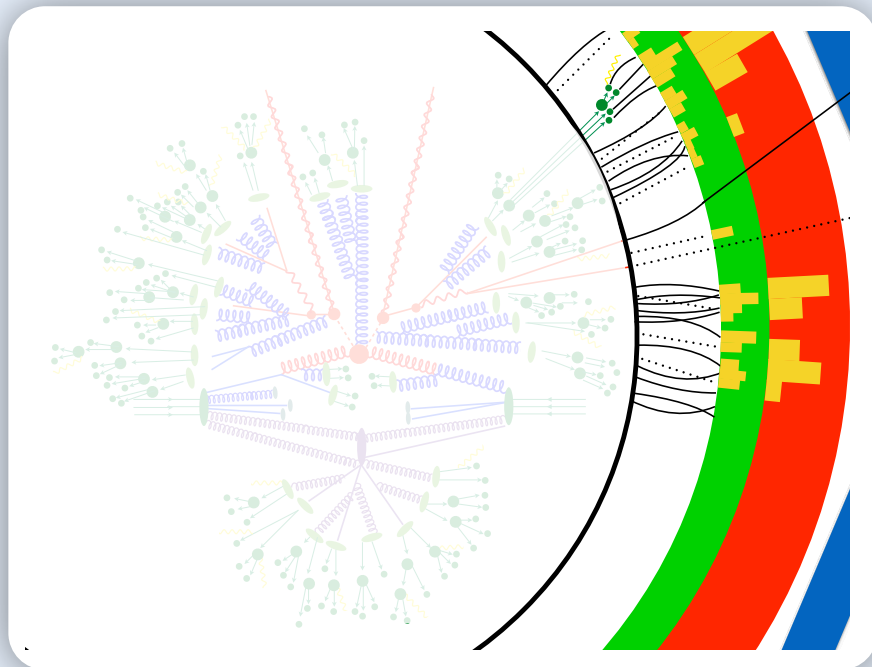
2205.03413

Part II: Unfolding

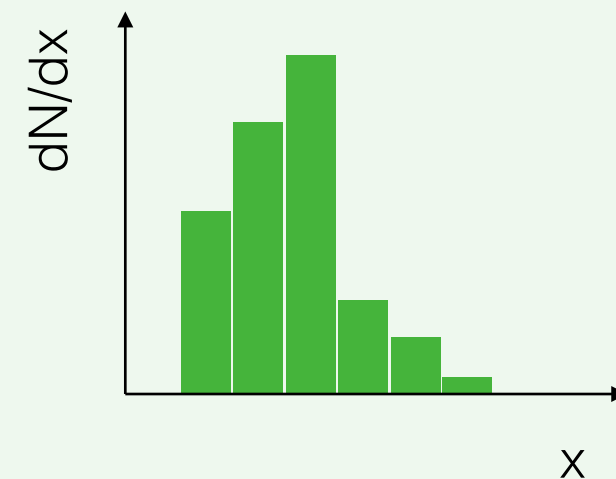


Detector
Level

Particle
Level



Traditional
Approach



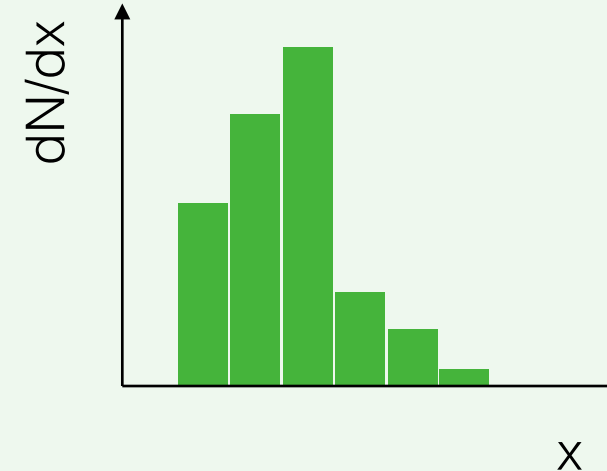
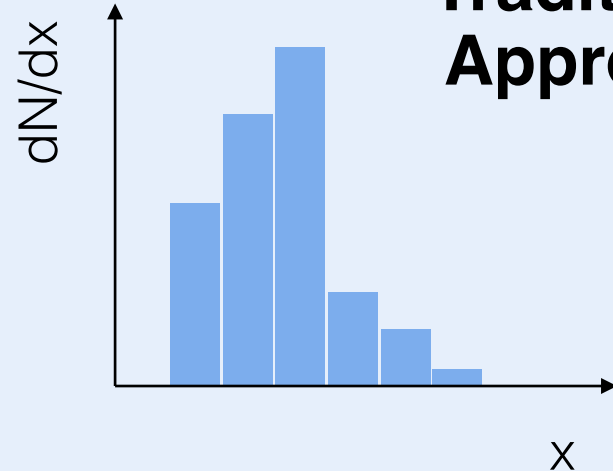
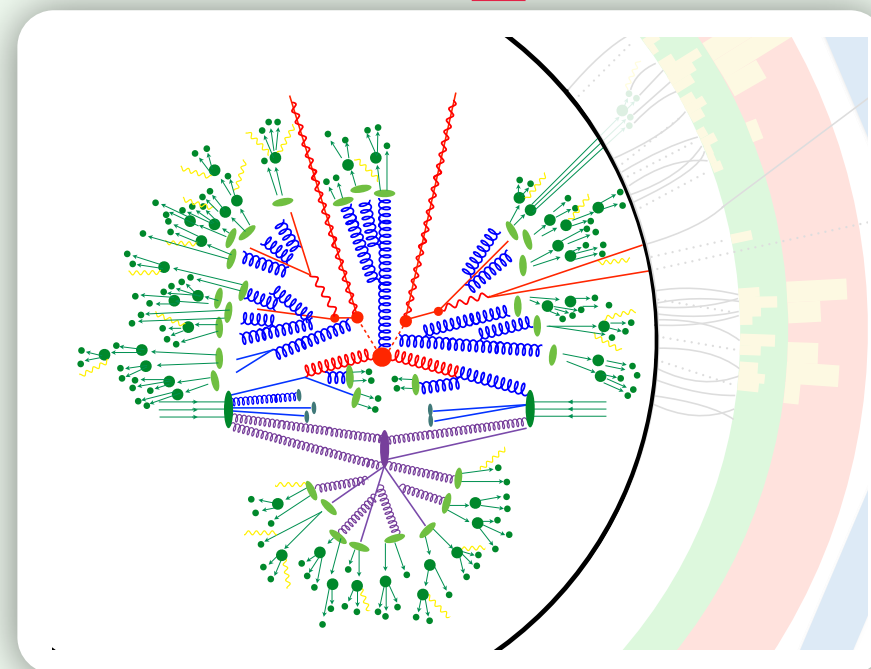
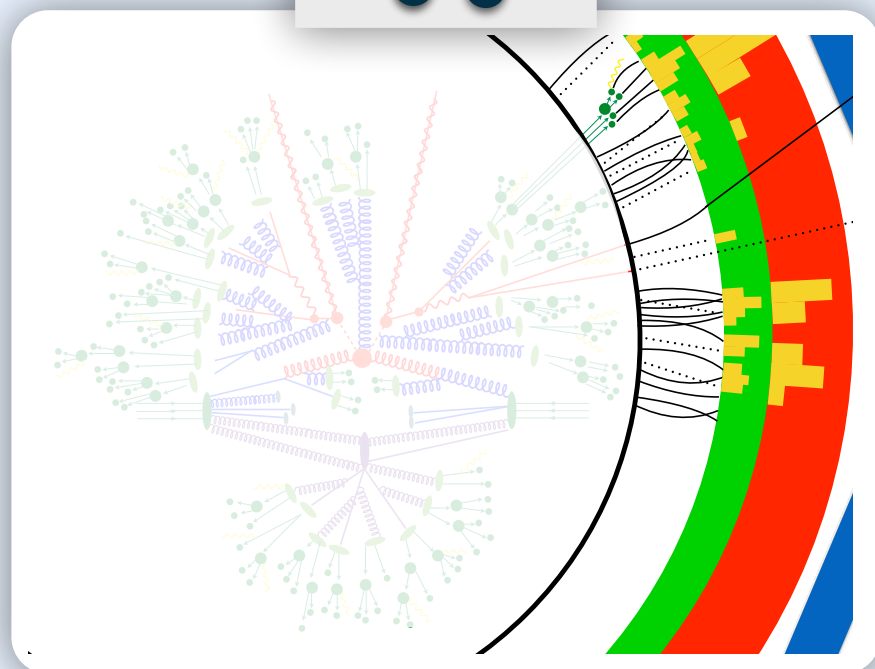
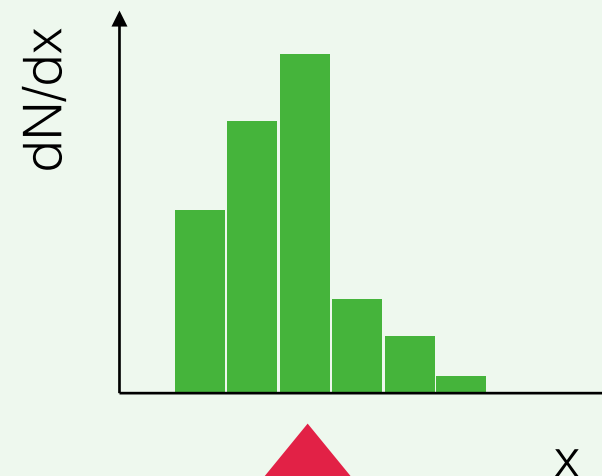
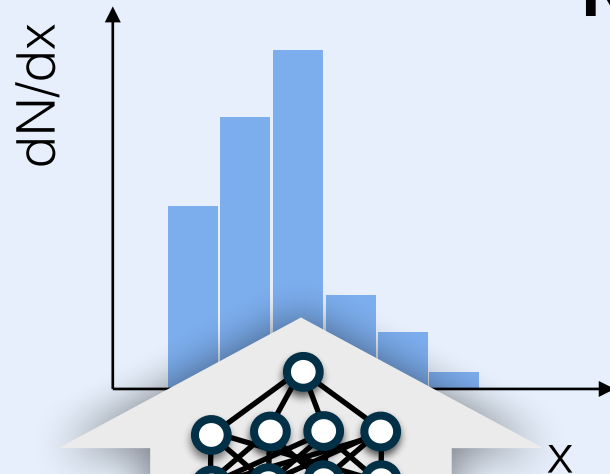
Target
observable

2203.16722

Detector
Level

NN reco tailored
for unfolding

Particle
Level

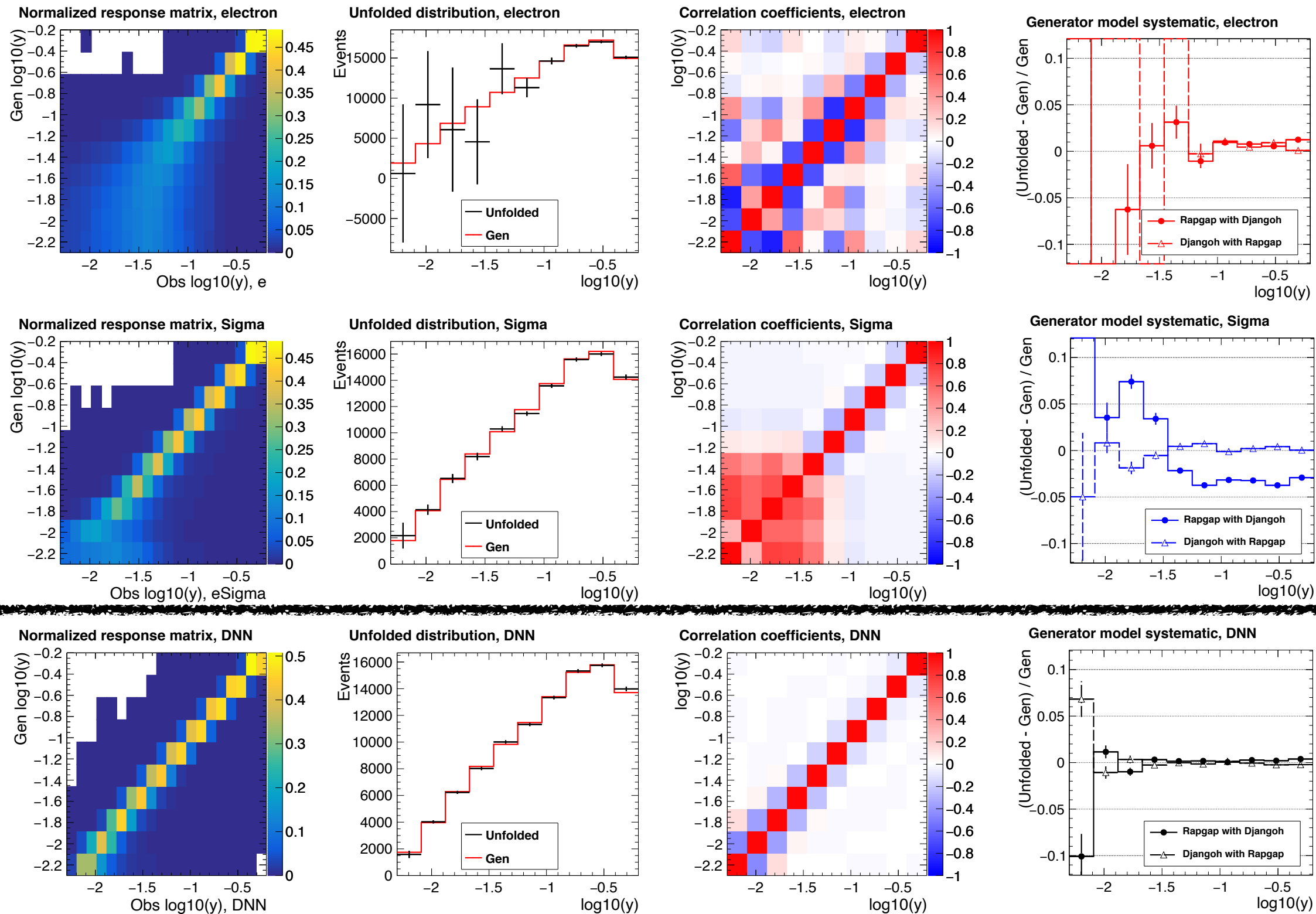


Traditional
Approach

Target
observable

Tailored Reco for Unfolding

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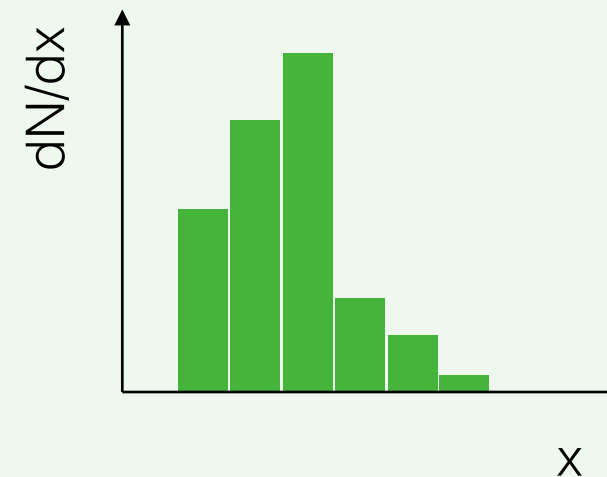
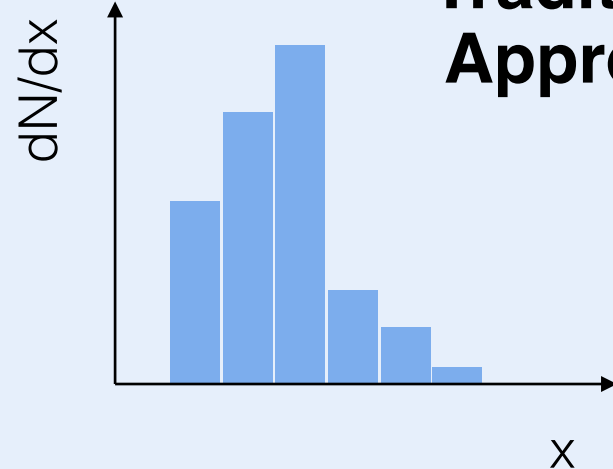
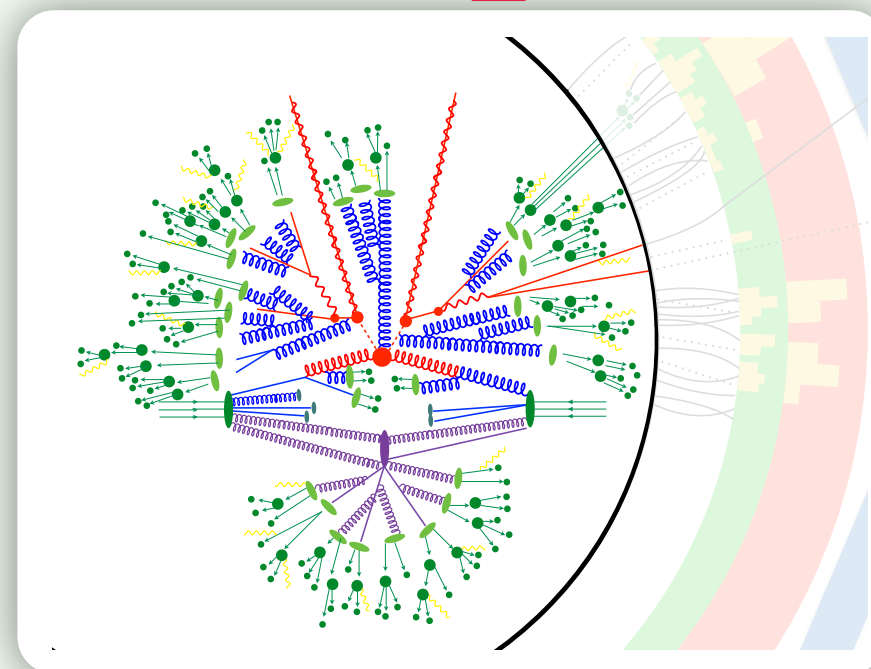
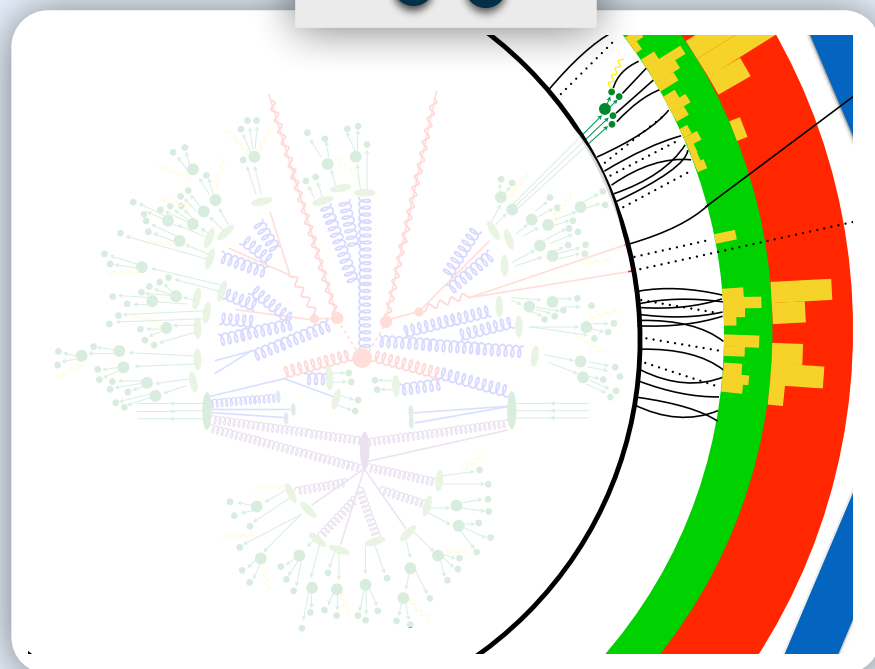
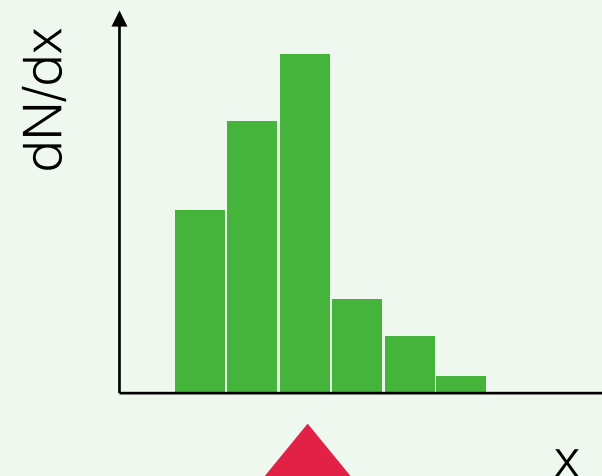
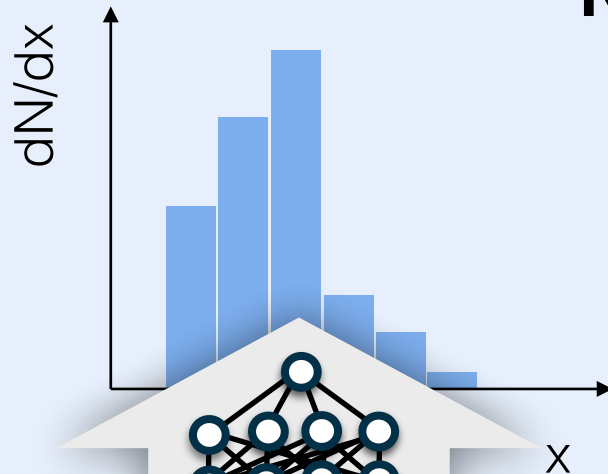
Impact on inclusive DIS: better stat + syst

2203.16722

Detector
Level

NN reco tailored
for unfolding

Particle
Level



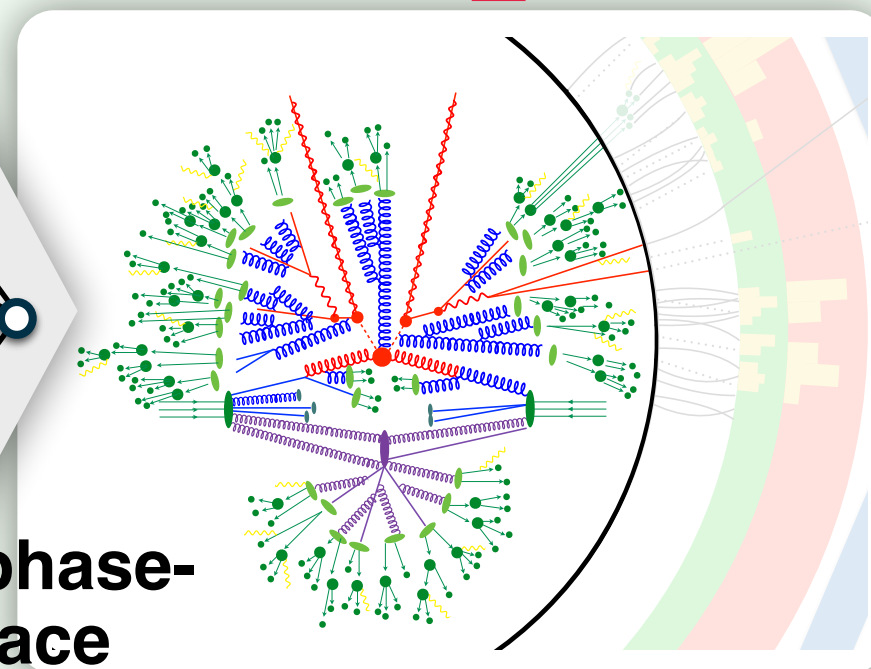
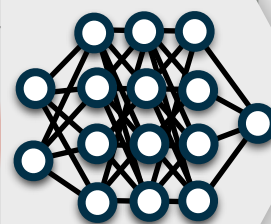
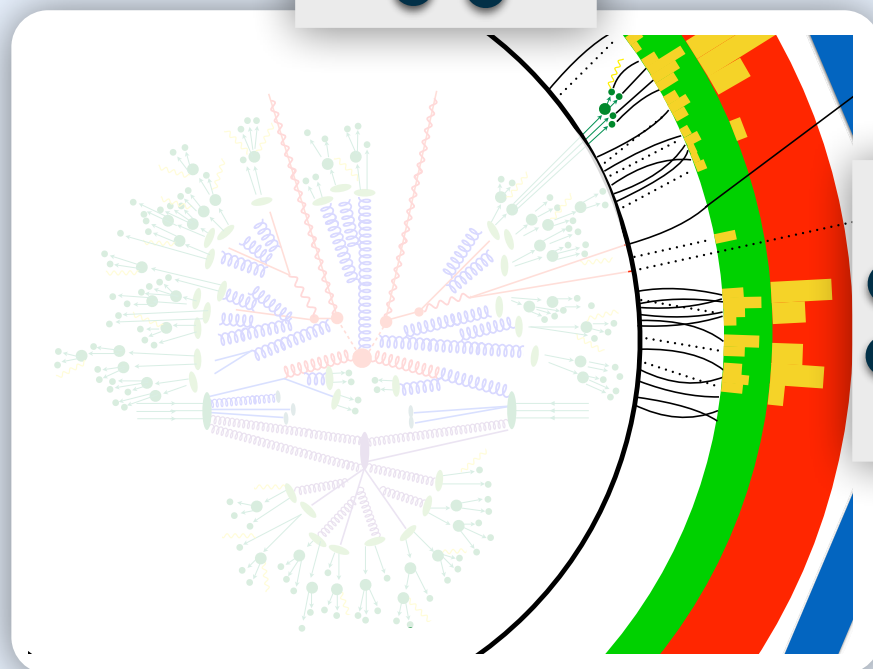
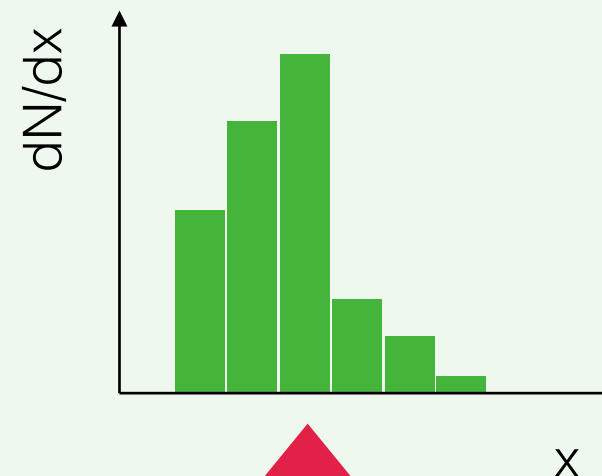
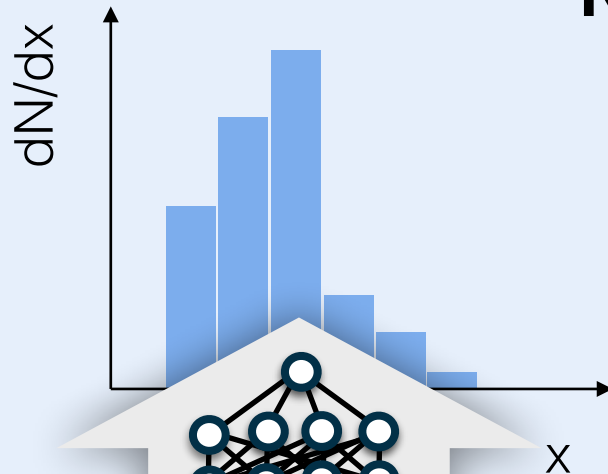
Traditional
Approach

Target
observable

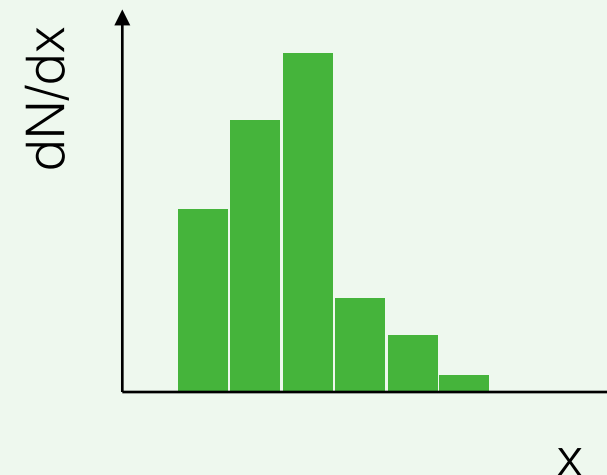
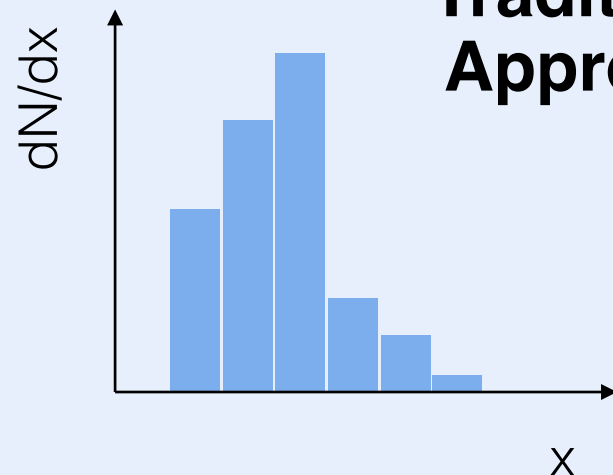
**Detector
Level**

**NN reco tailored
for unfolding**

**Particle
Level**



**Full phase-
space
unfolding**

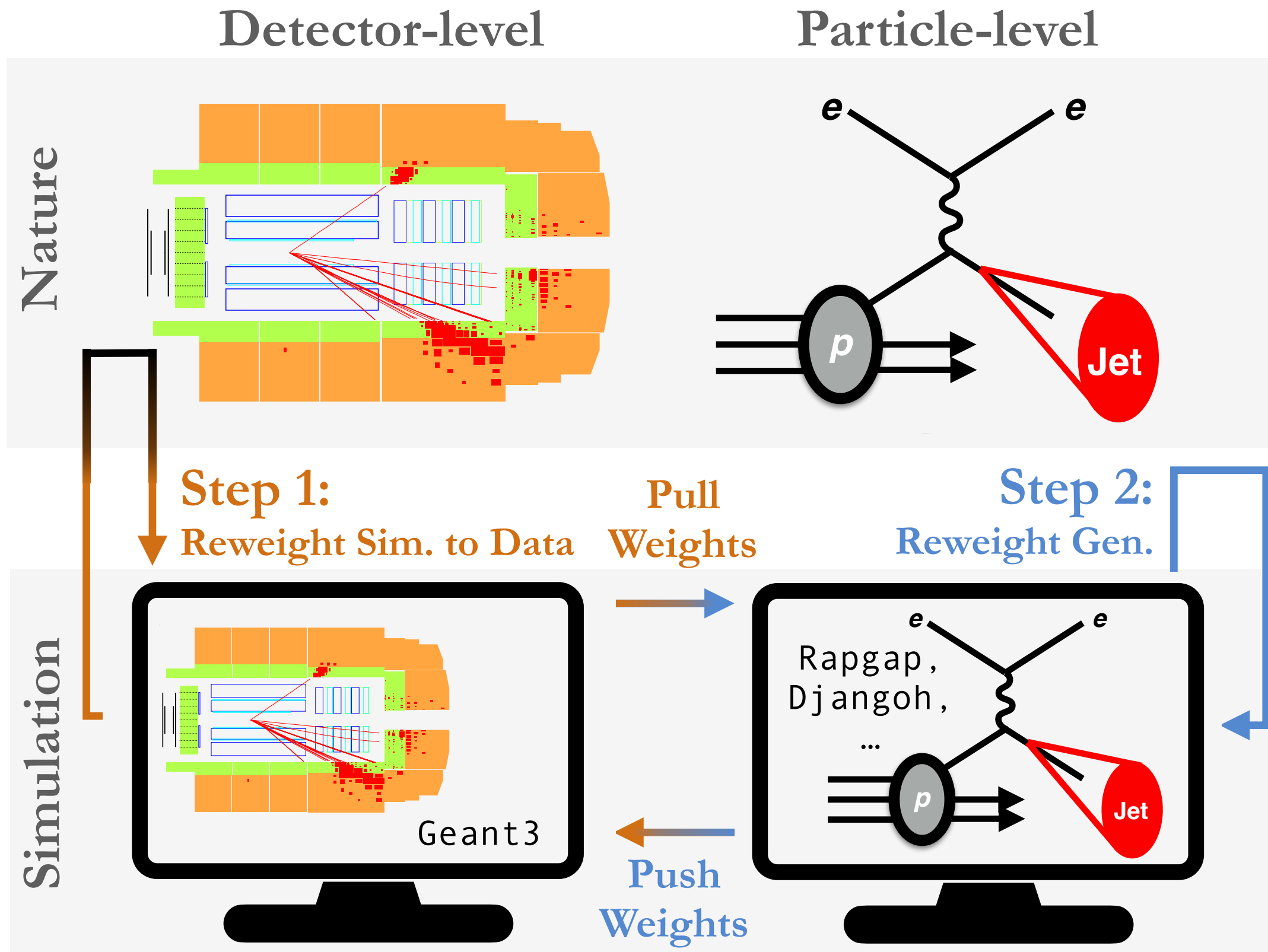


**Traditional
Approach**

**Target
observable**

Full Phase via OmniFold

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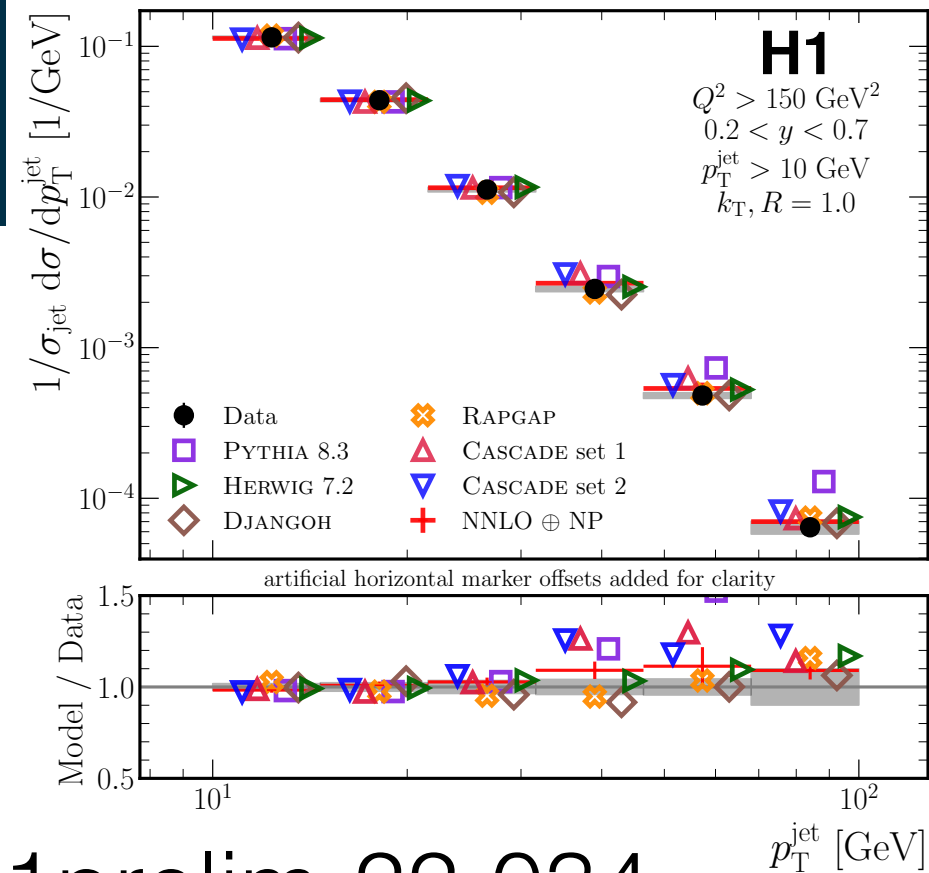
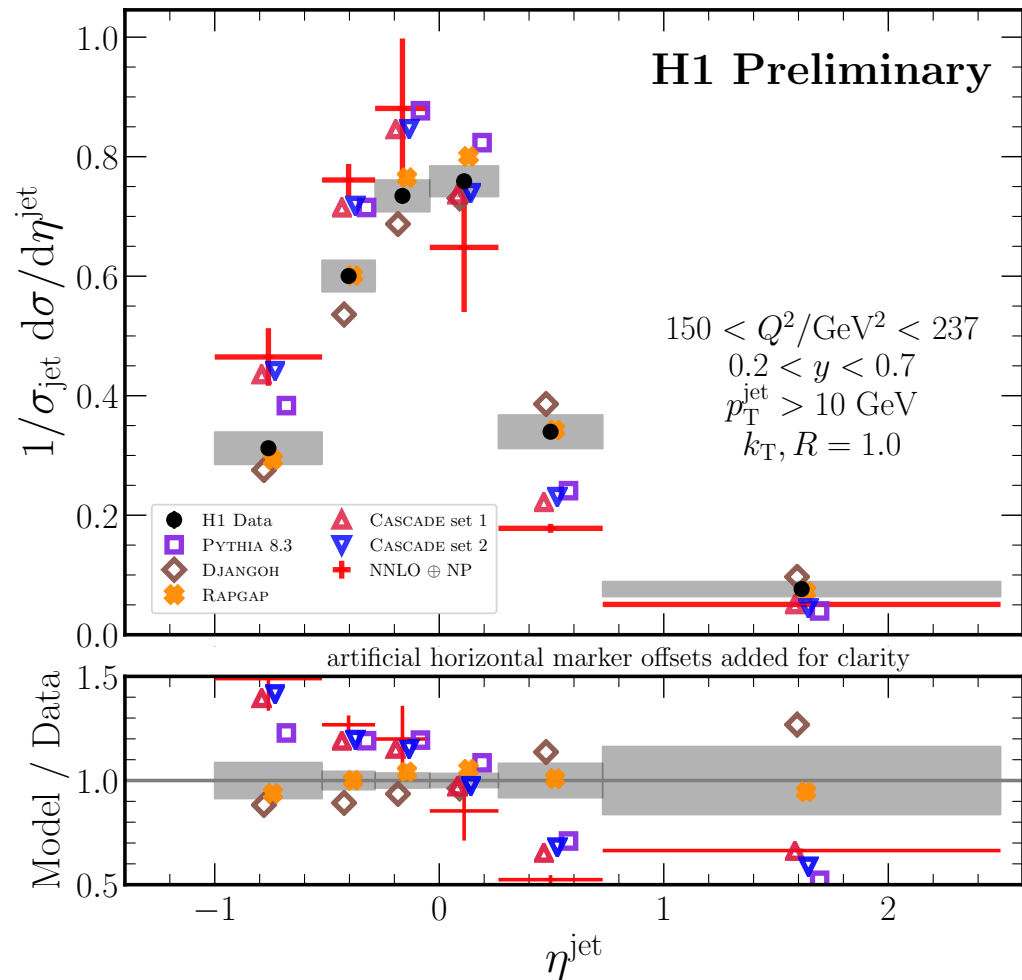


This is not a dream!

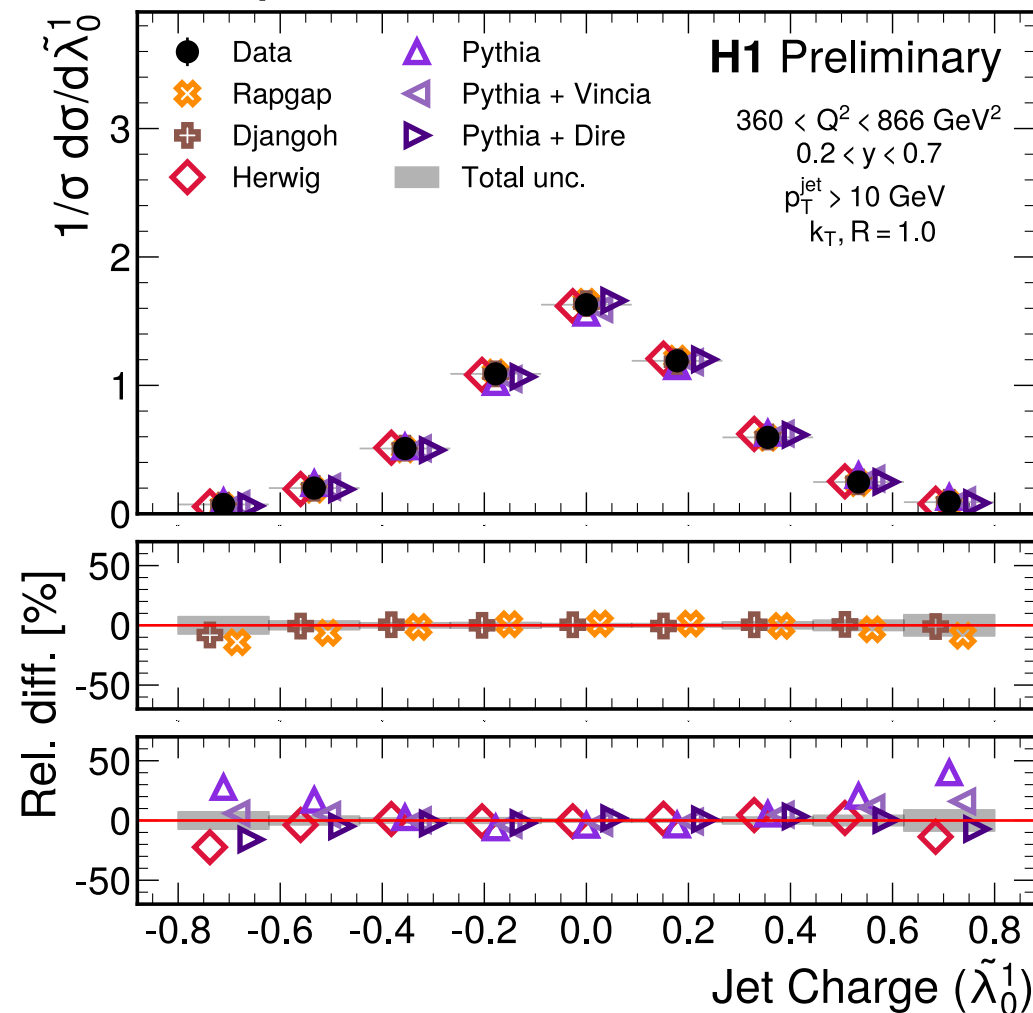
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26

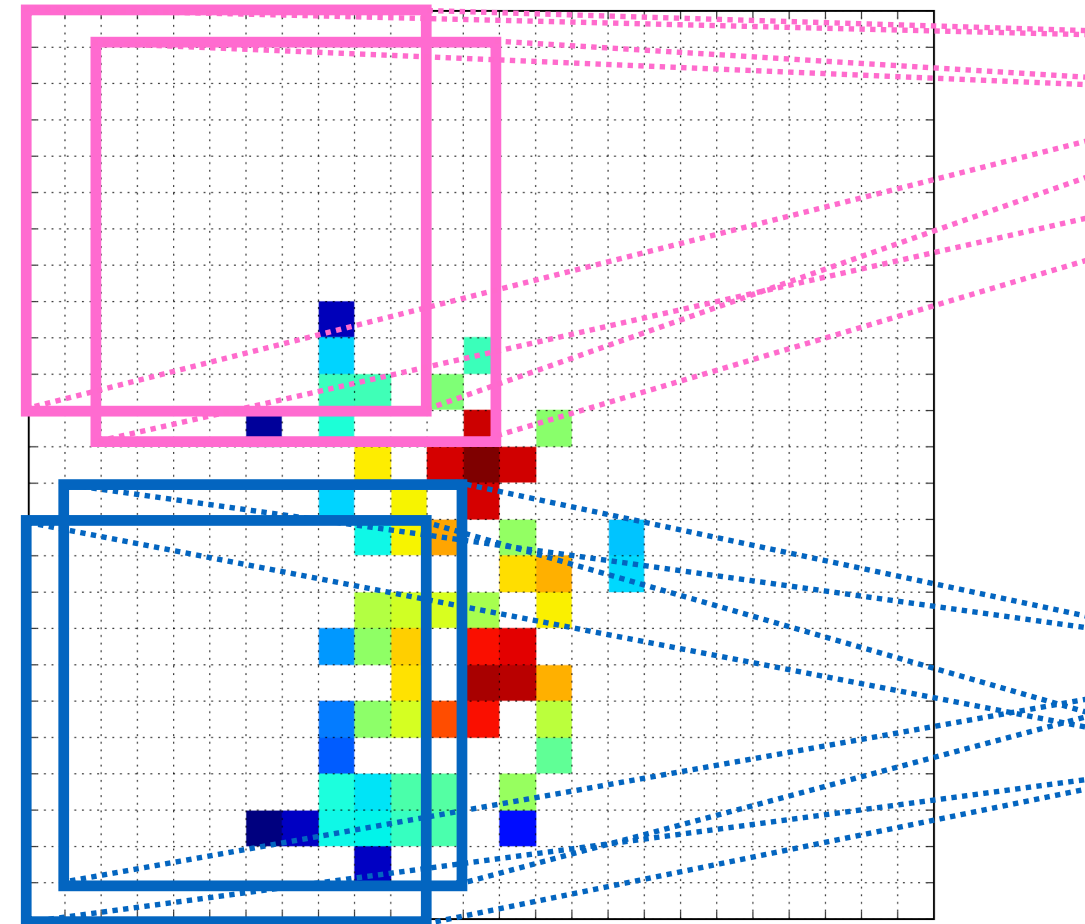
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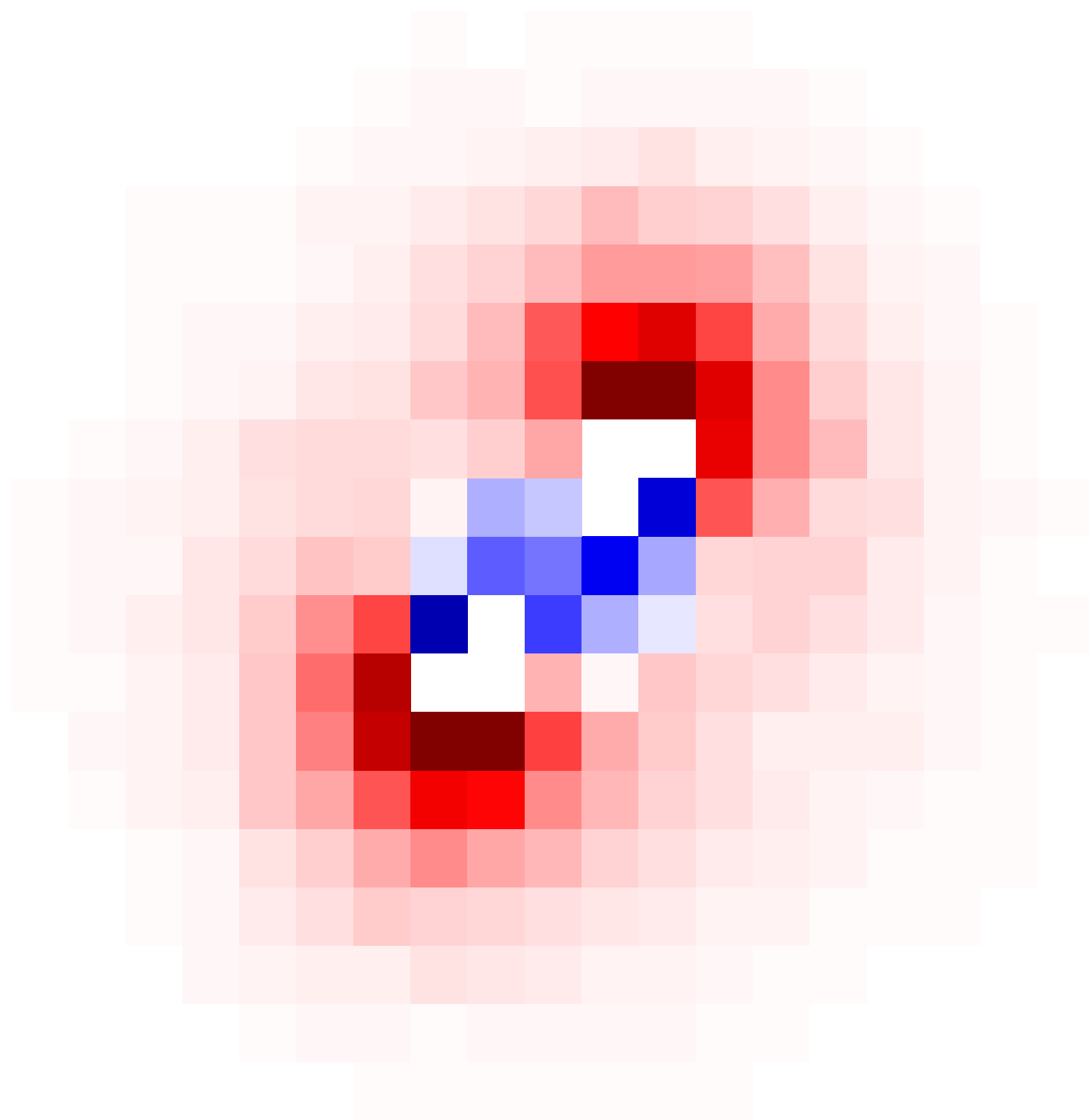
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AI/ML has a great potential to enhance reconstruction and analysis at a future ep collider



We can take advantage of developments from pp and already start exploring applications at HERA/EIC



Fin.