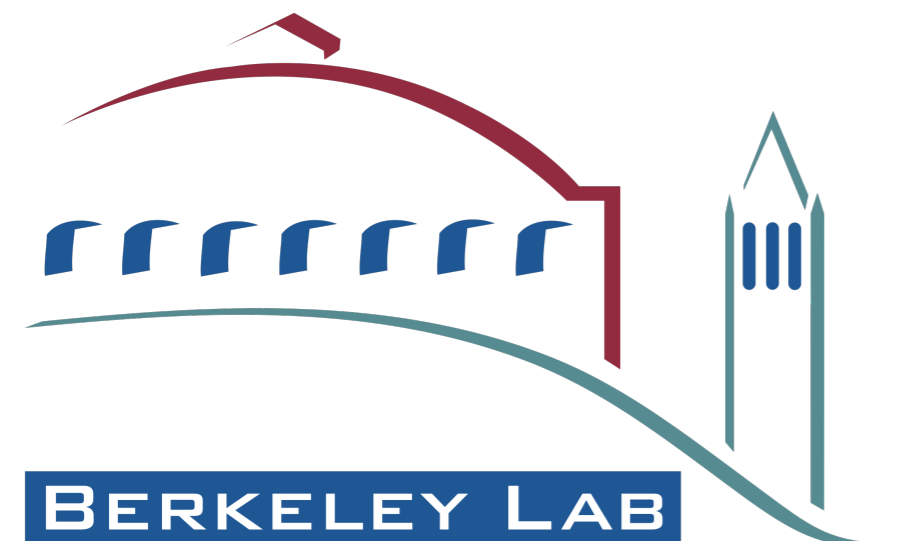


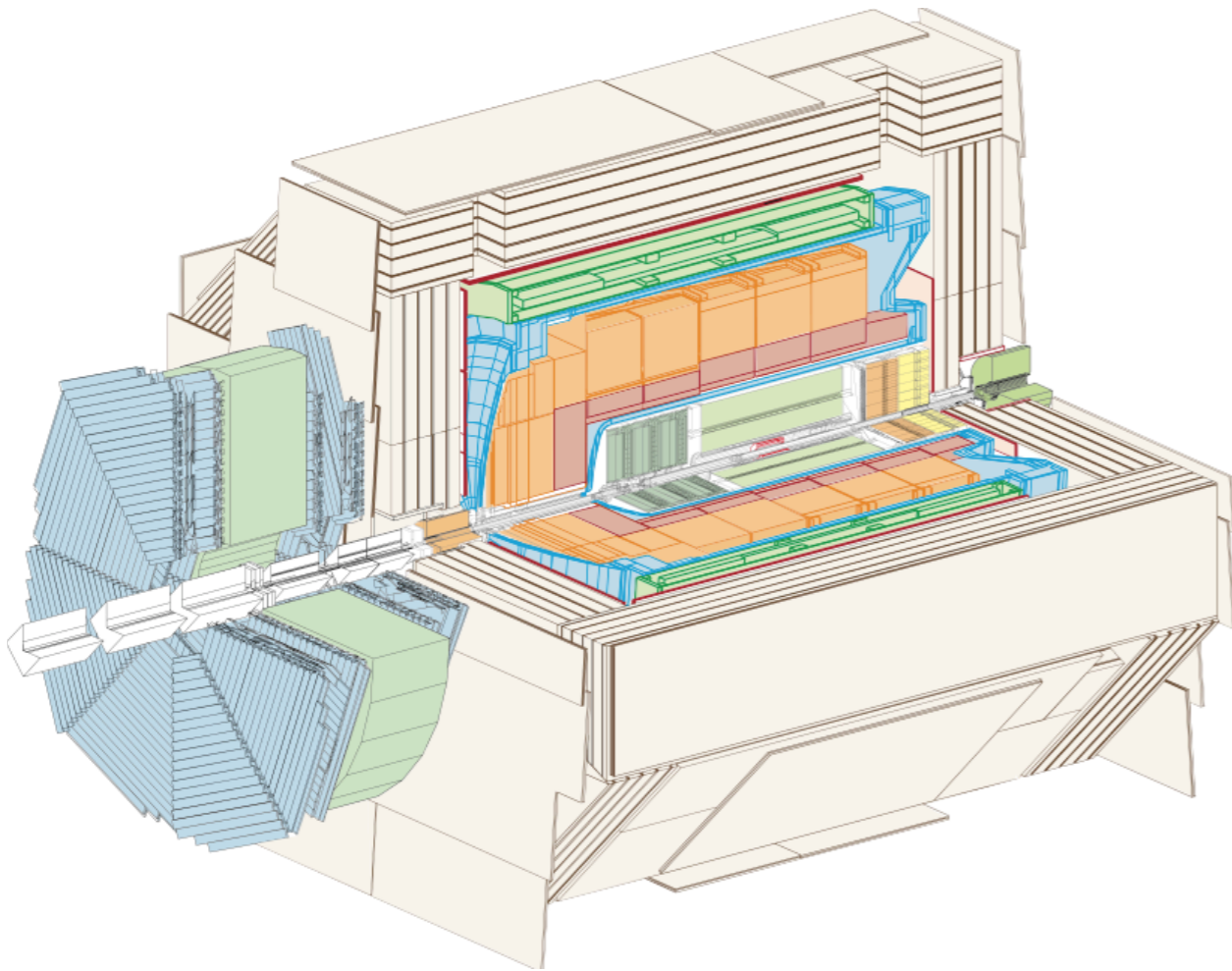
Unfolded Measurements With H1 using MultiFold

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on behalf of the H1 Collaboration



H1 at HERA



- H1 Detector at the positron-proton collider, HERA. Hosted in Hamburg Germany
- Major goal was to study internal structure of the proton through deep inelastic scattering

$$e(k) + q(p_1) \rightarrow e'(k_\ell) + jet(k_J) + X$$

Quick Overview

- Lots of discussion on unfolding multi-differential cross sections
 - A bit more!
- Let's also look at a previously inaccessible observable
 - Moments requiring Un-binned unfolding

H1 Data

- Same data / selection / unfolding as [arXiv:2108.12376](https://arxiv.org/abs/2108.12376)
 - “Measurement of lepton-jet correlation in deep-inelastic scattering with the H1 detector using machine learning for unfolding”

- H1 Data from 2006 and 2007 periods at 130 pb^{-1}
 - $\sqrt{s} = 320 \text{ GeV}$, eP collisions



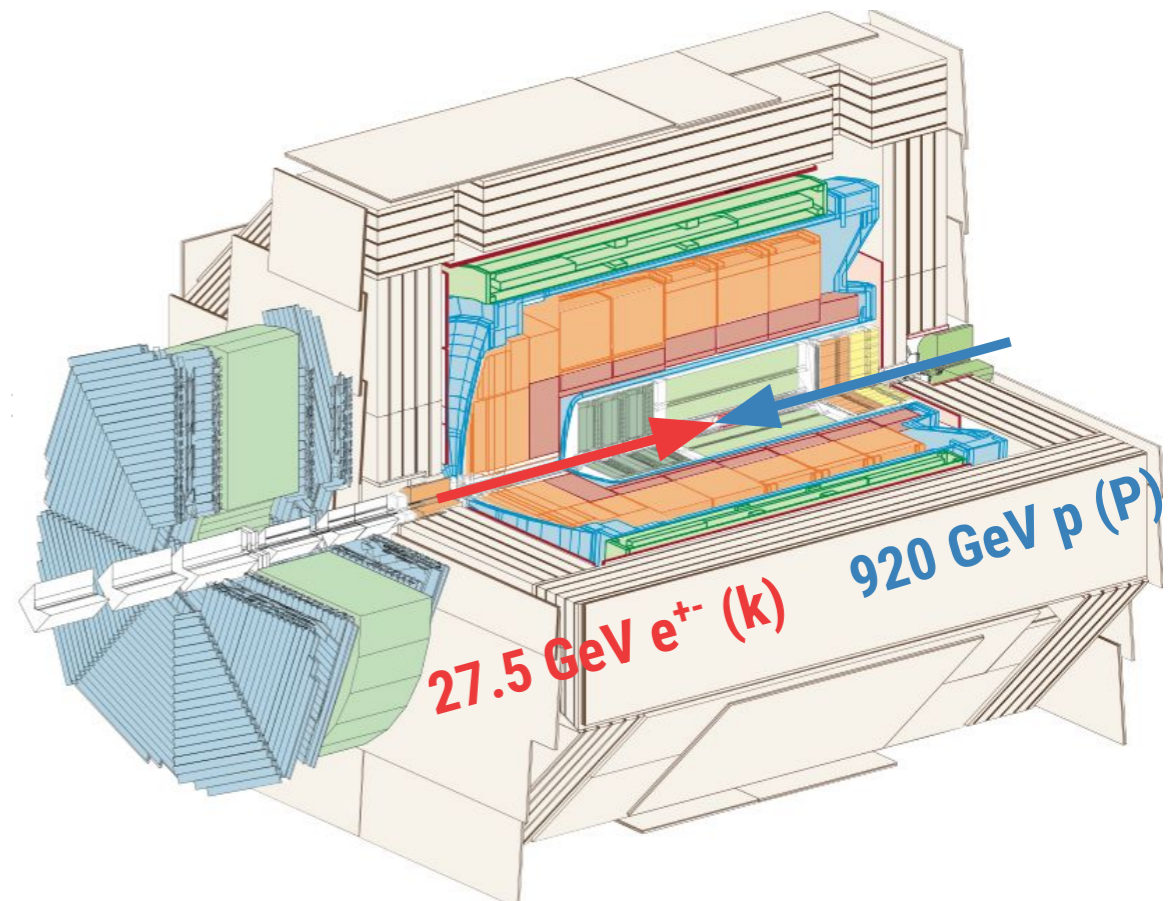
- Fiducial Cuts:

- $0.2 < y < 0.7$
- $Q^2 > 150 \text{ GeV}^2$
- $p_T^{\text{jet}} > 10 \text{ GeV}$
- $-1 < \eta_{\text{lab}} < 2.5$
- $k_T, R = 1.0$
- $q_{\perp}/p_{T,\text{jet}} < 0.3$

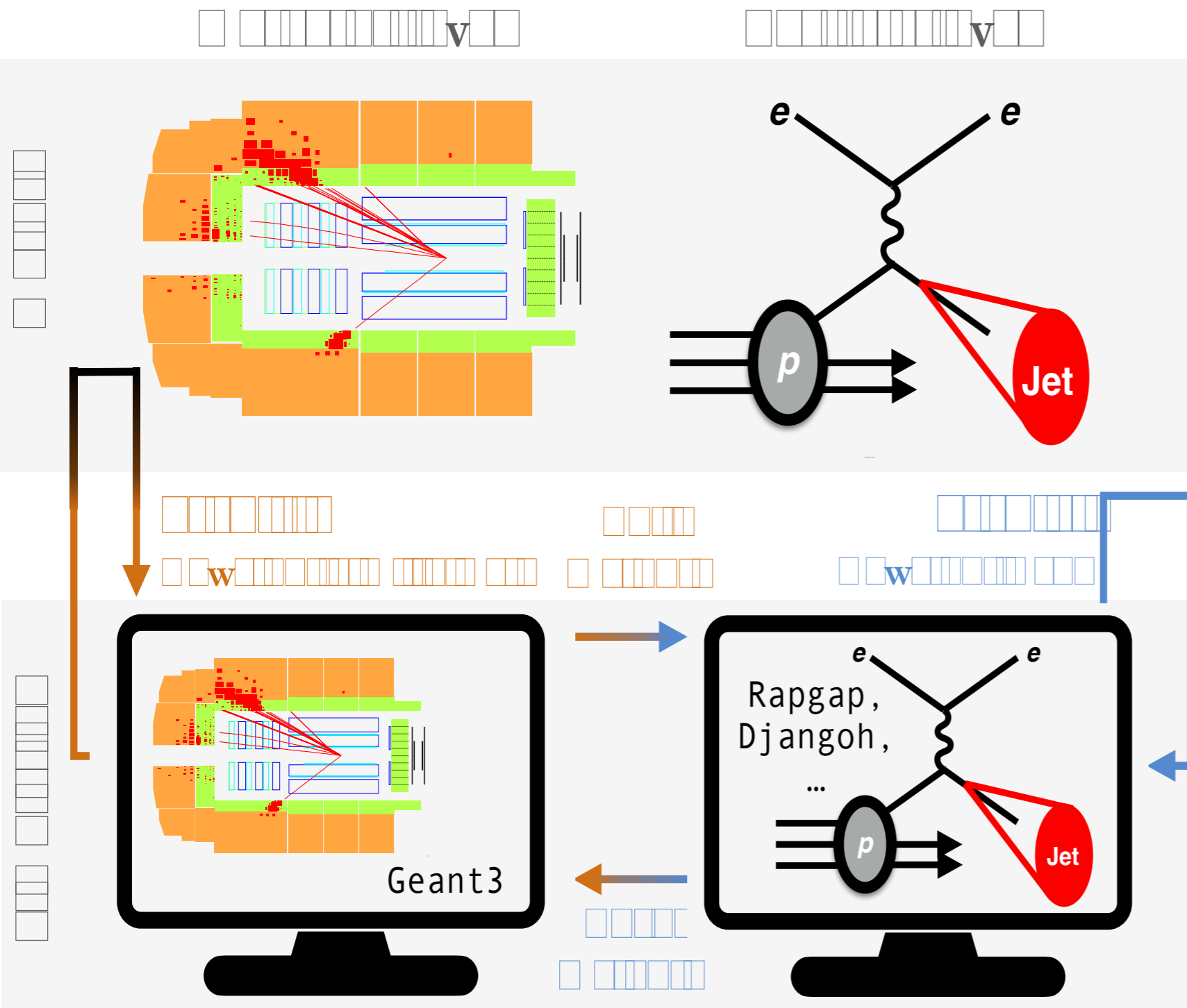
For asymmetry Measurement:

Cut on $q_{\perp}/p_{T,\text{jet}}$ to satisfy $P_{\perp} \gg q_{\perp}$:

$$p_{T,\text{jet}} \approx P_{\perp}/2$$



OmniFold



2 step iterative approach

1. Events from detector level sim. are reweighted to match the data
2. Create a "new simulation" by transforming weights to a proper function of the generated events

Classifiers used to approximate 2 likelihood functions:

1. reco MC to Data reweighting
2. Previous and new Gen reweighting

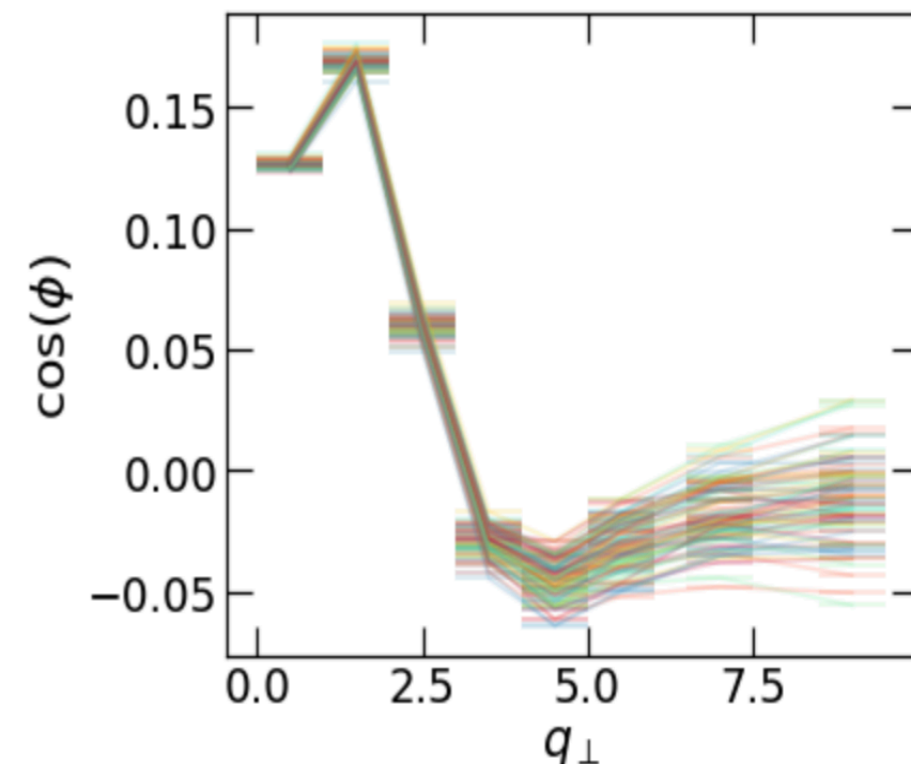
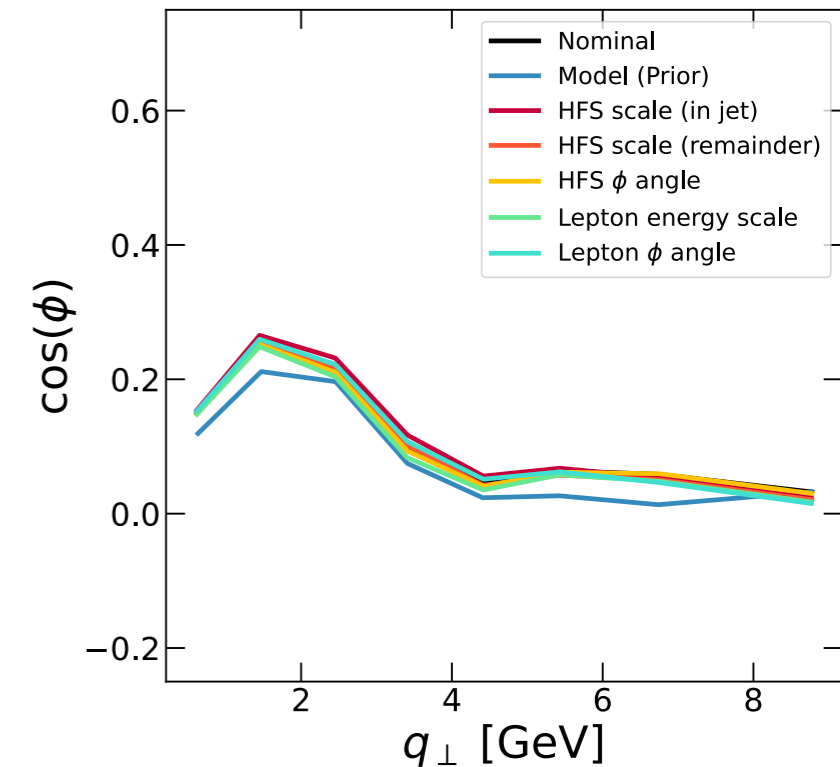
Uncertainties

Systematic uncertainties

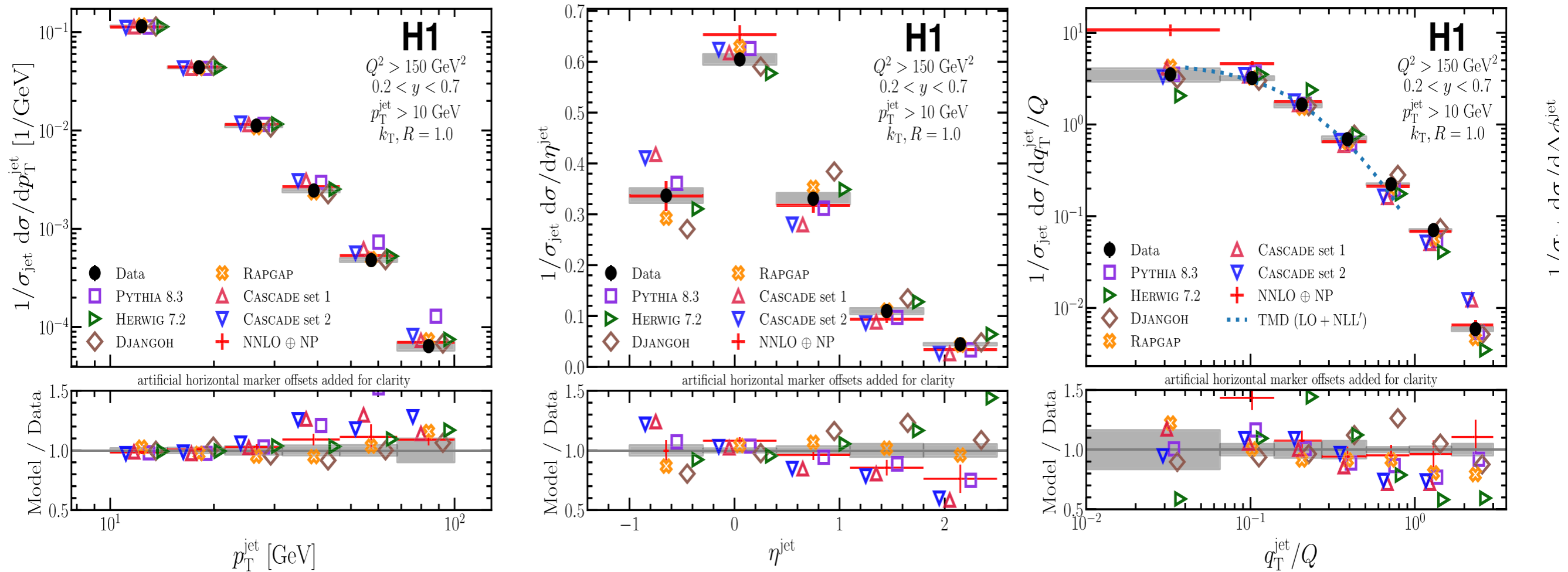
- **HFS energy scale:** $\pm 1\%$
- **HFS azimuthal angle:** ± 20 mrad
- **Lepton energy:** $\pm 0.5\%$ (mainly affects Q^2)
- **Lepton azimuthal angle:** ± 1 mrad (mainly affects Q^2)
- **Model uncertainty:** differences in unfolded results between Djangoh and Rapgap
- **QED uncertainty:** Use the variation of measured quantities when radiation is turned off in the simulation

Statistical Uncertainty

- **Bootstrapping**
- Each event is given a new initial weight
- ~ 100 bootstraps
- Repeat entire unfolding process



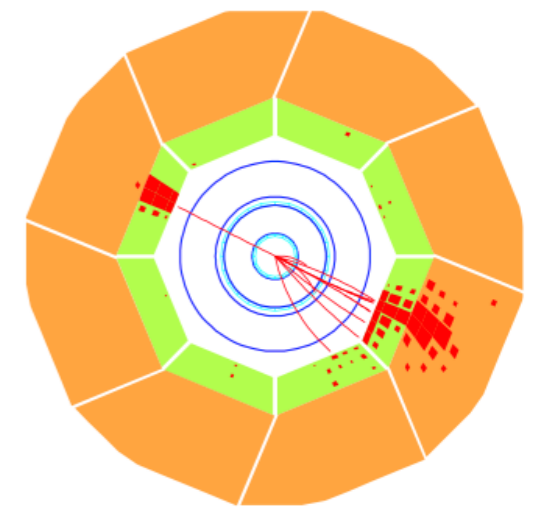
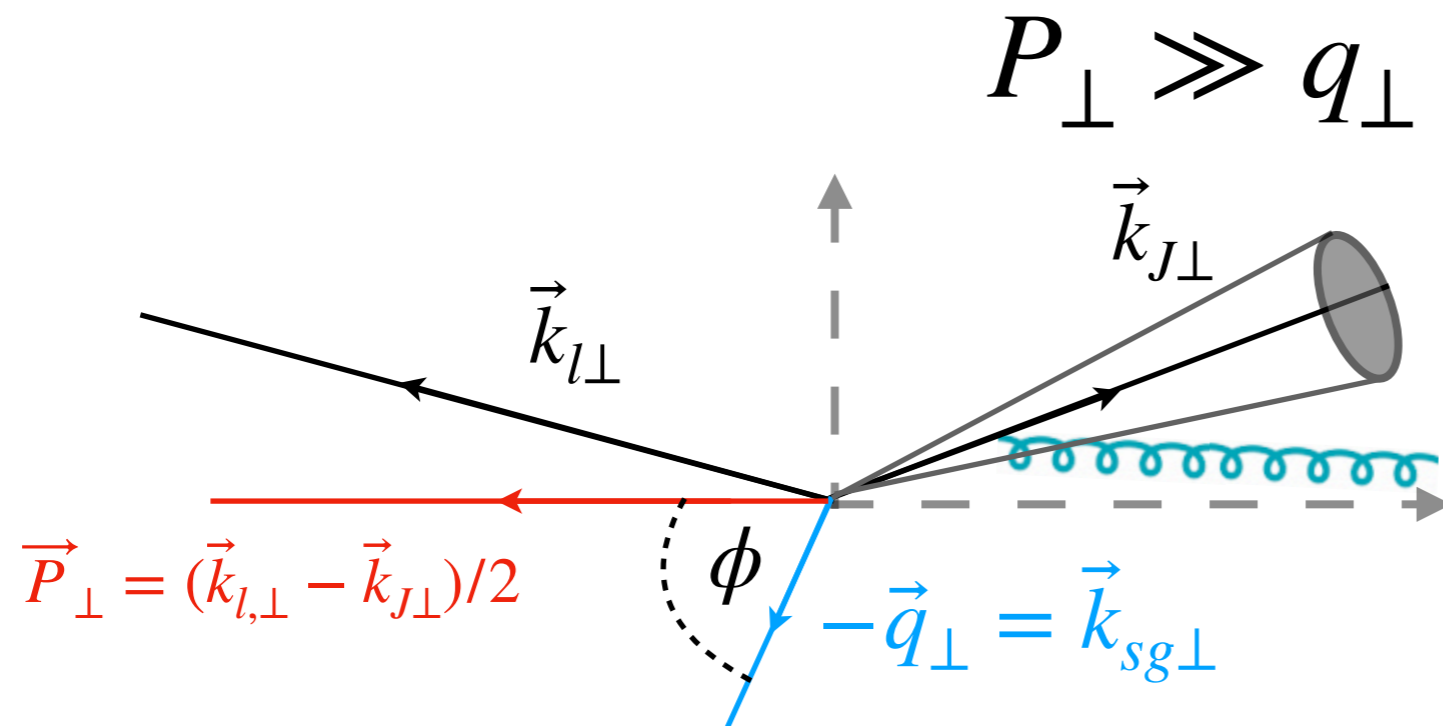
H1 Differential Cross sections



- Production cross sections as a function of p_T^{jet} , η^{jet} , and jet momentum balance q_T^{jet}/Q
- Simultaneous unfolding of eight observables:
 - $p_x^e, p_y^e, p_z^e, p_T^{\text{jet}}, \eta^{\text{jet}}, \phi^{\text{jet}}, \Delta\phi^{\text{jet}}, q_T^{\text{jet}}/Q$

Lepton Jet Asymmetry

- Total transverse momentum of the outgoing system $\vec{q}_\perp = \vec{k}_{\ell\perp} + \vec{k}_{J\perp}$, is typically *small* but *nonzero*
- Imbalance can come from perturbative initial and final state radiation
 - e.g. Emission of soft gluon with momentum $k_{\perp g}$
 - unrelated to TMDs or intrinsic transverse momentum of target gluons
- Depending on kinematics, soft gluon radiation can dominate
 - Radiative corrections enhanced approximately as $(\alpha_s \ln^2 P_\perp^2 / q_\perp^2)^n$



$$e(k) + q(p_1) \rightarrow e'(k_\ell) + jet(k_J) + X$$

Lepton Jet Asymmetry

Key Ingredients:

$$\vec{q}_\perp = \vec{k}_{\ell\perp} + \vec{k}_{J\perp}$$

$$\vec{P}_\perp = (\vec{k}_{\ell\perp} - \vec{k}_{J\perp}) / 2$$

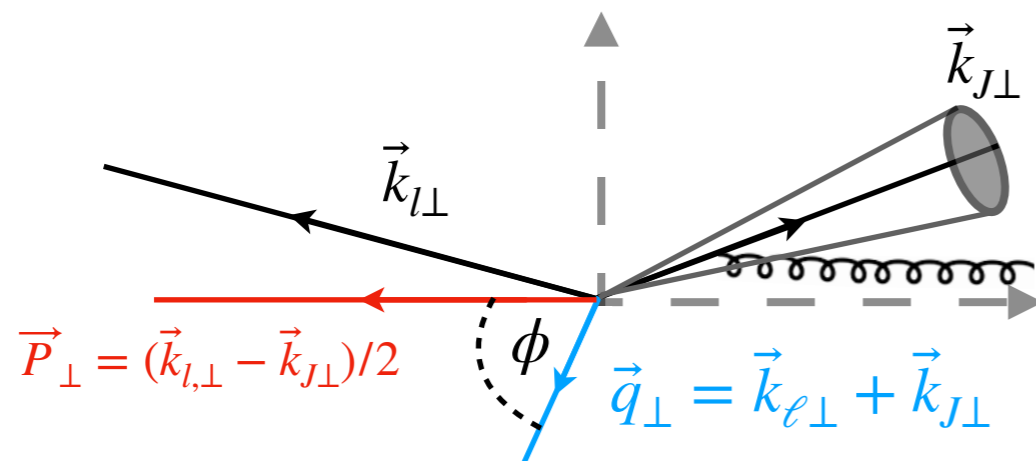
$$\phi = \text{acos}[(\vec{q}_\perp \cdot \vec{P}_\perp) / (q_\perp P_\perp)]$$

- $q_\perp = \text{Total transverse momentum}$

- $P_\perp = \text{Transverse momentum difference}$

- $\phi = \text{Angle between } q_\perp \text{ and } P_\perp$

Final Observable:
 $\langle \cos(n\phi) \rangle$ for $n = 1, 2, 3$



Multifold used to unfold:

$$p_x^e, p_y^e, p_z^e, p_T^{\text{jet}}, \eta^{\text{jet}}, \phi^{\text{jet}}, \Delta\phi^{\text{jet}}, q_T^{\text{jet}}/Q$$

Momentum conservation:

$$\vec{q}_\perp = -\sum_i^{\text{soft}} \vec{k}_{i\perp}$$

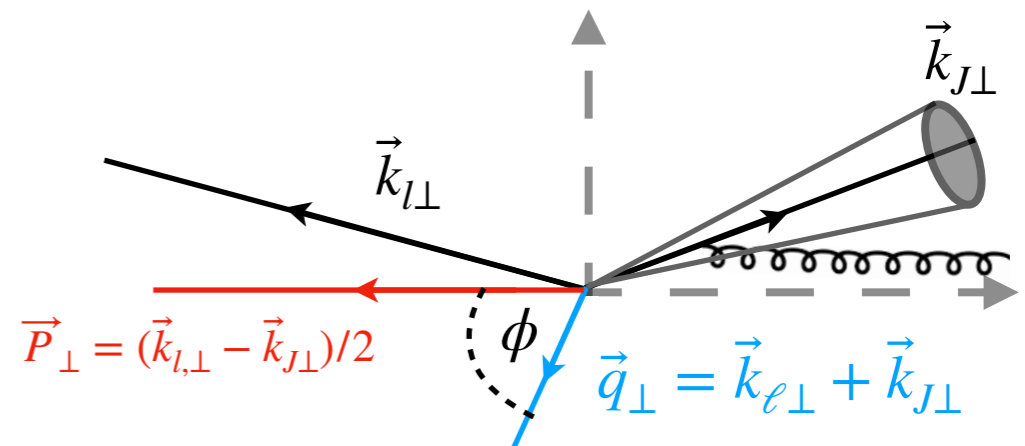
Motivation

1. Probes soft gluon radiation $S(g)$
 - Soft gluon radiation can be the primary contribution to asymmetry
 - [10.1103/PhysRevD.104.054037](https://arxiv.org/abs/10.1103/PhysRevD.104.054037)
2. Asymmetry is perturbative
 - Opportunity to compare to unfolded H1 data
3. May represent a vital reference for other signals, in particular TMD PDF measurements
 - Factorize contributions TMD PDFs and Soft gluon radiation
4. Observable is sensitive to gluon saturation phenomena, possibly measurable at the EIC
 - [10.1103/PhysRevLett.130.151902](https://arxiv.org/abs/10.1103/PhysRevLett.130.151902)

Putting it Together

$$\phi = \text{acos}[(\vec{q}_\perp \cdot \vec{P}_\perp) / |\vec{q}_\perp| |\vec{P}_\perp|]$$

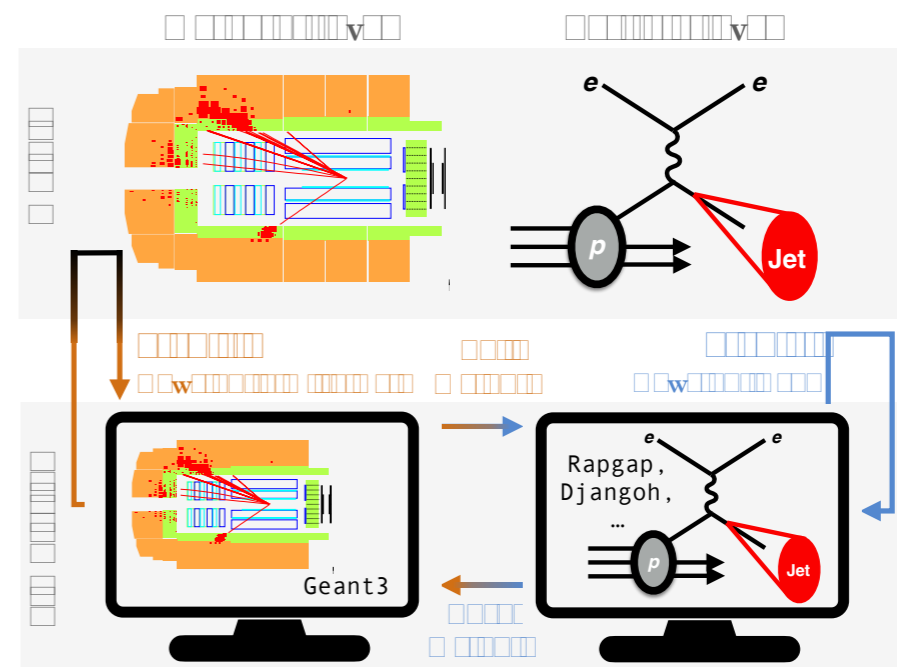
1. Obtain the azimuthal asymmetry angle, ϕ , in each event
2. Obtain unfolding event weight from MultiFold Step 2, ω_i , for each event, i



Calculate
 $\langle \cos(n\phi) \rangle$ for $n = 1, 2, 3$

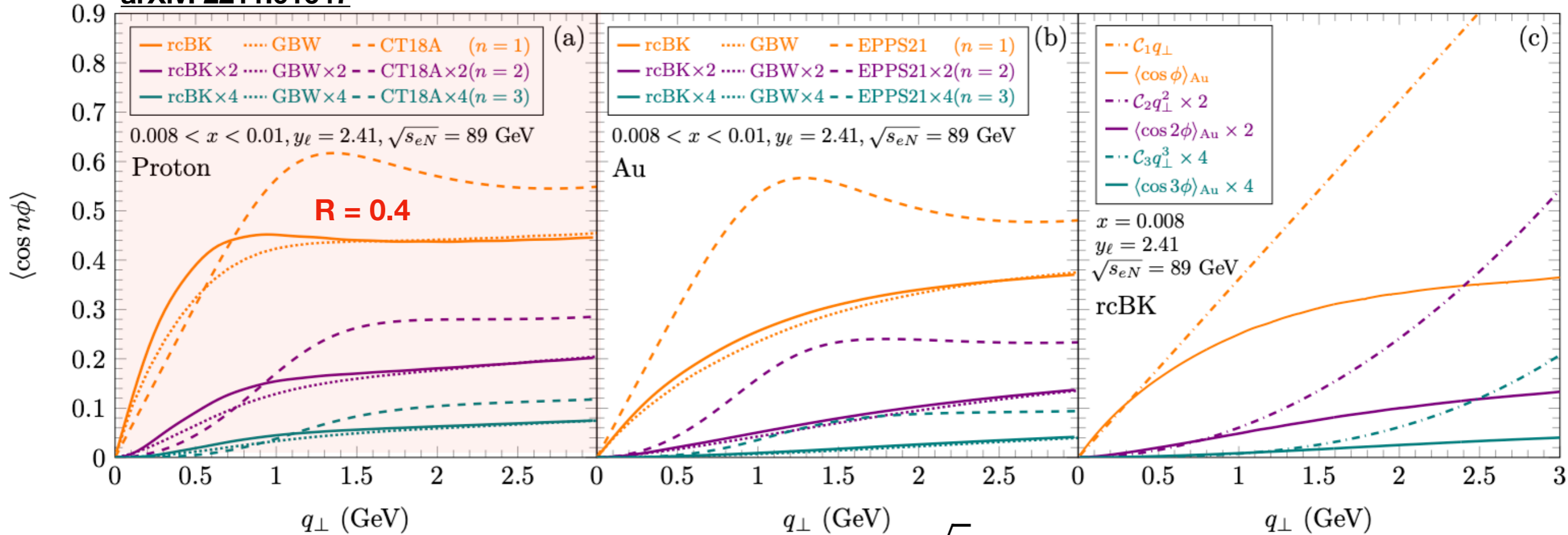
Report in bins of q_\perp GeV/c

$$\frac{\sum_i \omega_i \cos(n\phi_i)}{\sum_i \omega_i} \text{ for } n = 1, 2, 3$$



EIC Calculation @ HERA kinematics

arXiv: 2211.01647



$$\vec{q}_{\perp} = \vec{k}_{\ell\perp} + \vec{k}_{J\perp}$$

$$\vec{P}_{\perp} = (\vec{k}_{\ell\perp} - \vec{k}_{J\perp}) / 2$$

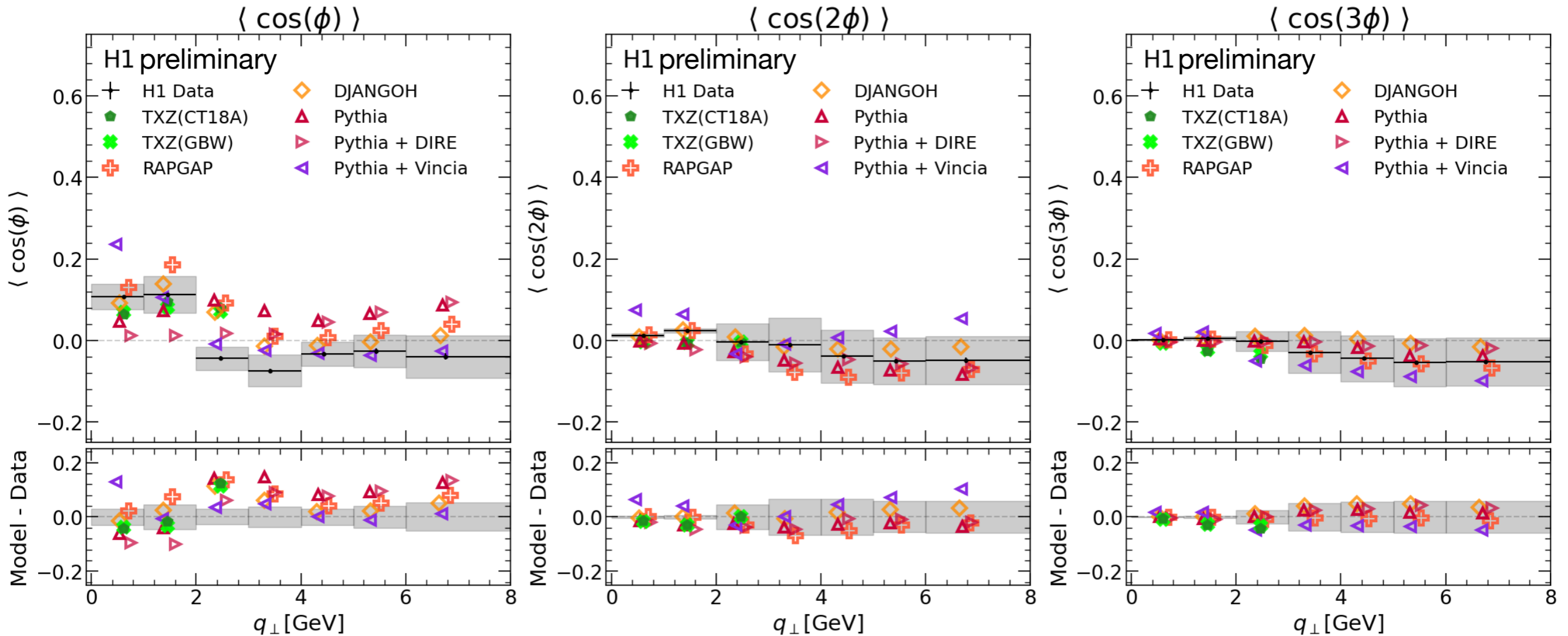
$$\sqrt{s} = 140 \text{ GeV}, P_{\perp} = 20 \text{ GeV},$$

$$y_l = 1.5, Q = 25 \text{ GeV}$$

Radiative corrections
 enhanced $\propto (\alpha_s \ln^2 P_{\perp}^2 / q_{\perp}^2)^n$

Plots above are for $R = 0.4$. Calculation done for this measurement w/ $R = 1.0$,
 Very good example of observable from ‘legacy’ dataset influencing future colliders
 Harmonics of saturation with the inputs GBW model and a TMD calculation CT18A PDF

Moments of Asymmetry Results



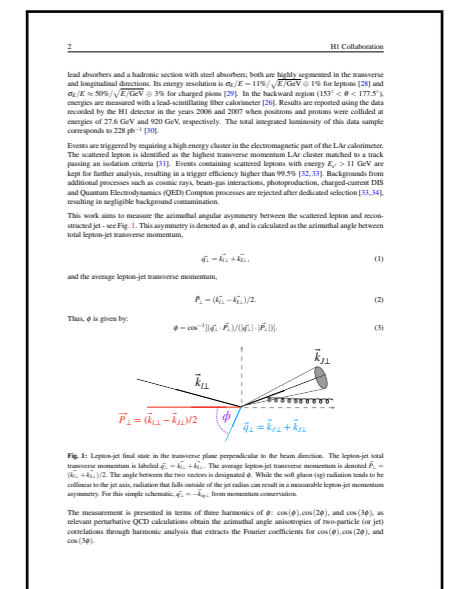
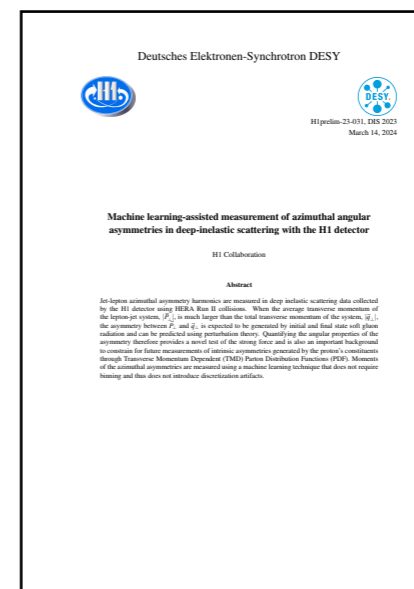
- Three harmonics of the azimuthal angular asymmetry between the lepton and leading jet as a function of q_{\perp} .
- Predictions from multiple simulations as well as a pQCD calculation are shown for comparison.
- Measurement indicates calculation may break down near $q_{\perp} \approx 3$ GeV

Conclusions

- Promising measurement to probe soft gluon radiation
 - Important reference for lepton-jet DIS measurements!
 - Comparisons with pQCD calculations, and 3 generators
 - Theory has qualitatively similar shape, but *underestimates contribution*
 - May point to larger non-perturbative contributions to this observable
- MultiFold
 - This work presents a measurement of *moments*, requiring the *un-binned unfolding!*
- H1 is a great example of exciting measurements using legacy datasets
 - First multidimensional un-binned unfolding using OmniFold
 - Novel observable with important implication for EIC
 - Simultaneous unfolding for Jet Substructure

<https://journals.aps.org/prl/pdf/10.1103/PhysRevLett.128.132002>

<https://doi.org/10.1016/j.physletb.2023.138101>



END

Backup

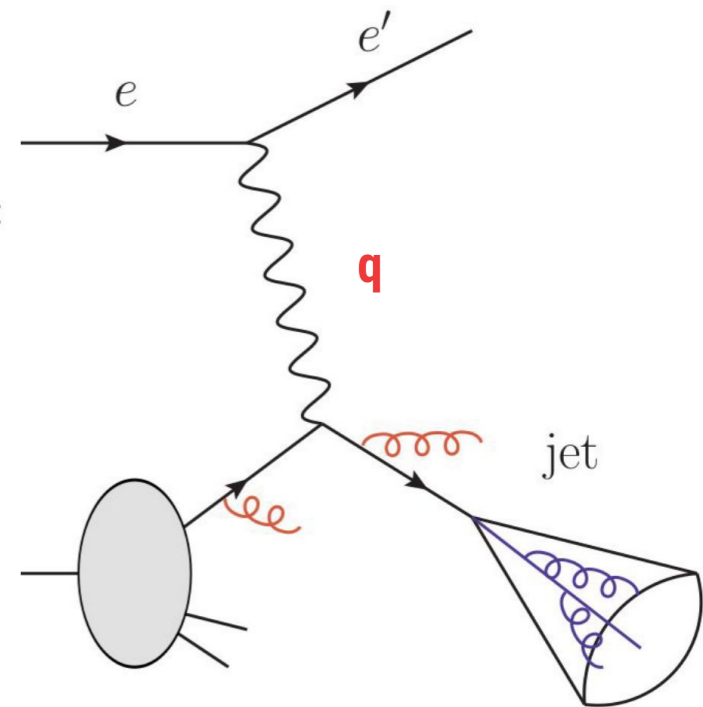
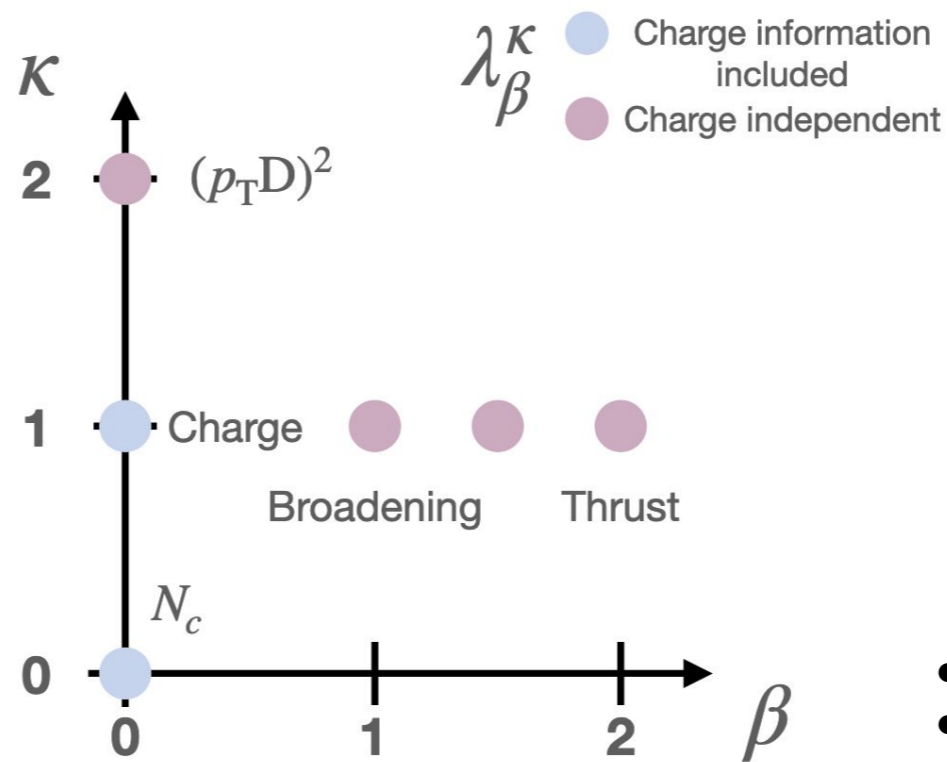
Jet Angularities

Use jet observables to study different properties of QCD physics:

- Infrared and collinear (IRC) safe λ_a^1 , $a = [0, 0.5, 1]$ and unsafe $\mathbf{p}_T \mathbf{D}$ angularities
- Charge dependent observables: \mathbf{Q}_j and \mathbf{N}_c
- Study the evolution of the observables with energy scale $Q^2 = -\mathbf{q}^2$

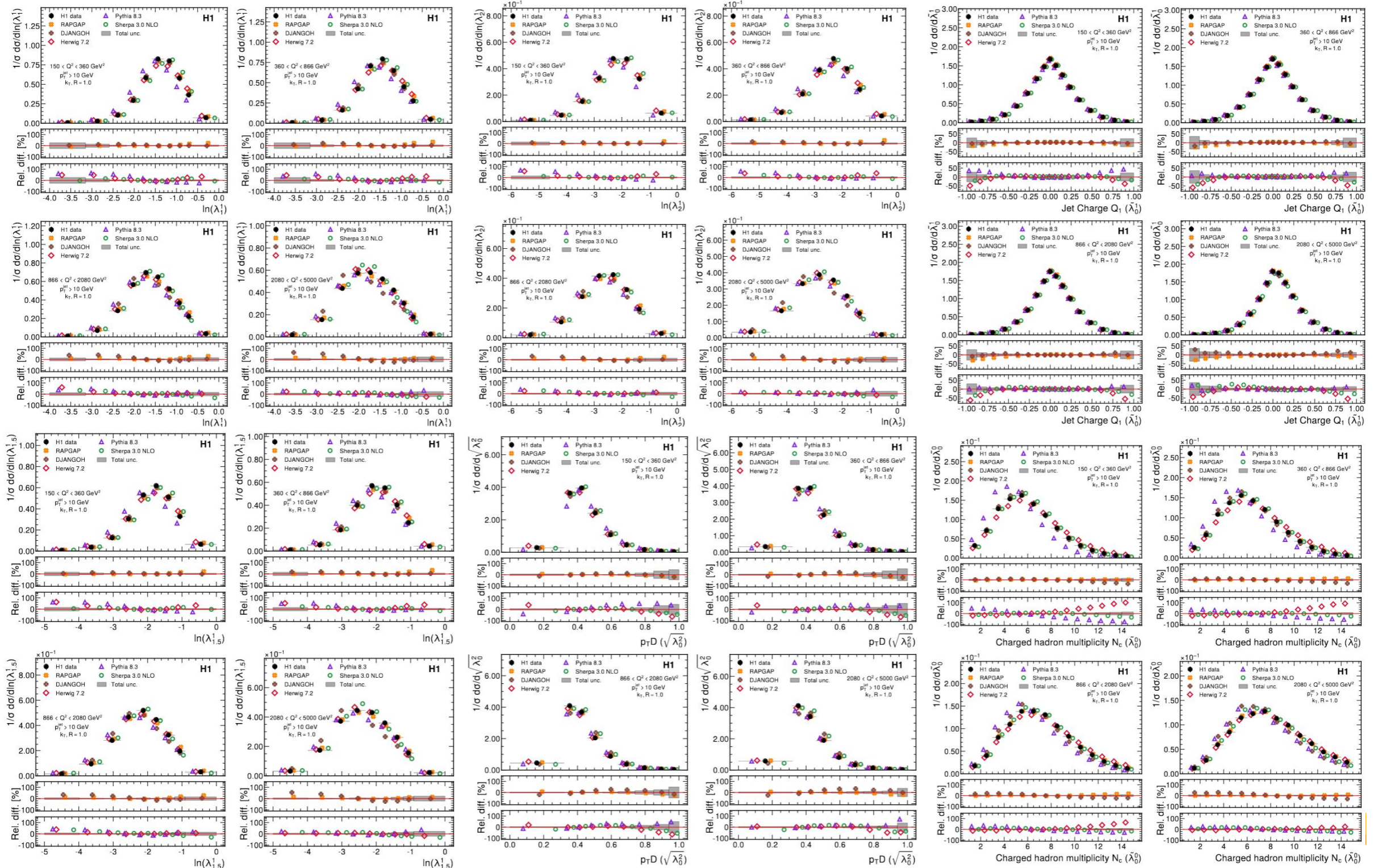
$$\lambda_{\beta}^{\kappa} = \sum_{i \in \text{jet}} z_i^{\kappa} \left(\frac{R_i}{R_0} \right)^{\beta}$$

$$\tilde{\lambda}_0^{\kappa} = Q_{\kappa} = \sum_{i \in \text{jet}} q_i \times z_i^{\kappa}.$$



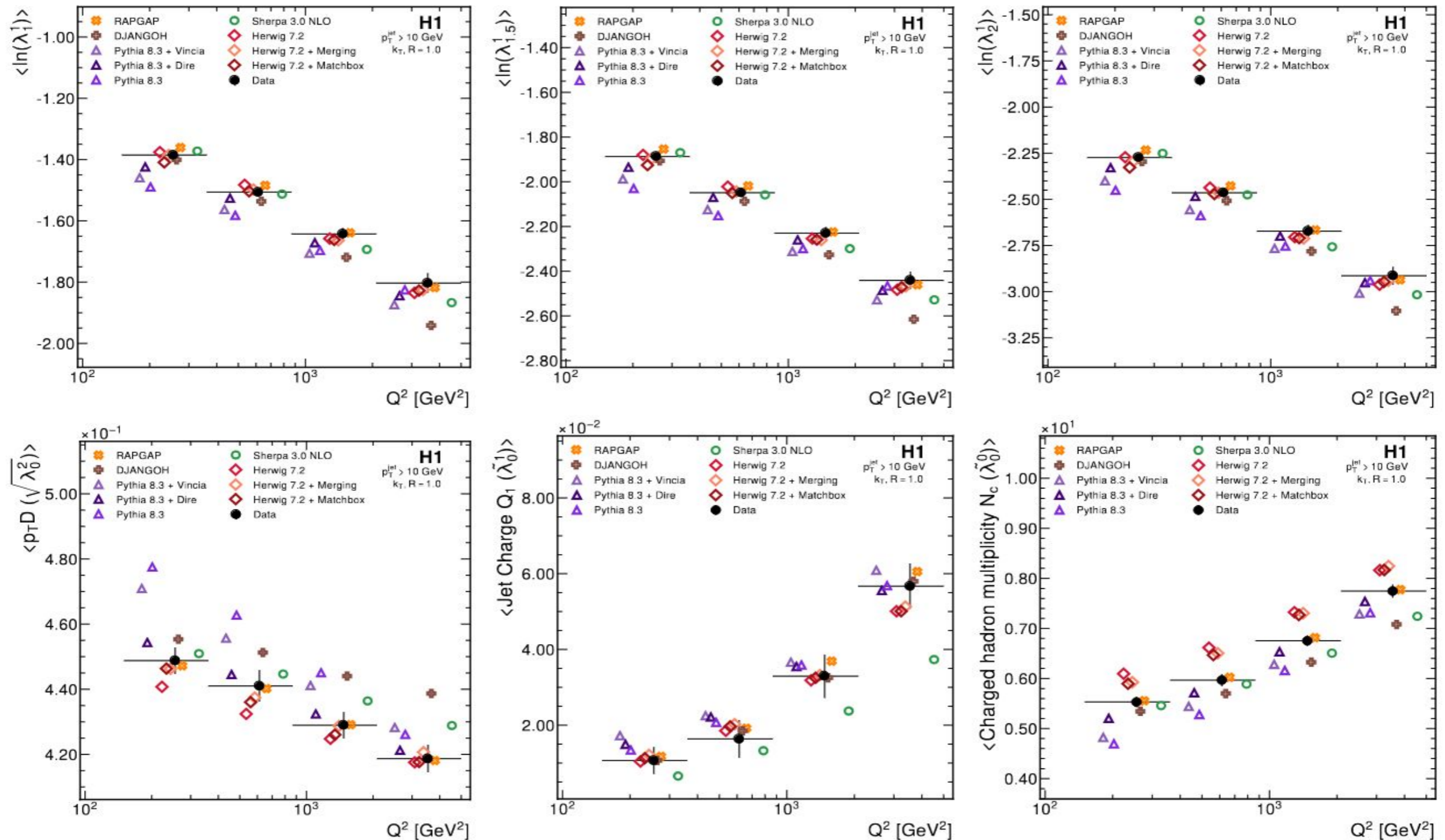
- z_i : longitudinal momentum fraction
- q_i : charge
- R_i : distance from jet axis in (eta, phi)

Multi-Differential H1 Jet Substructure



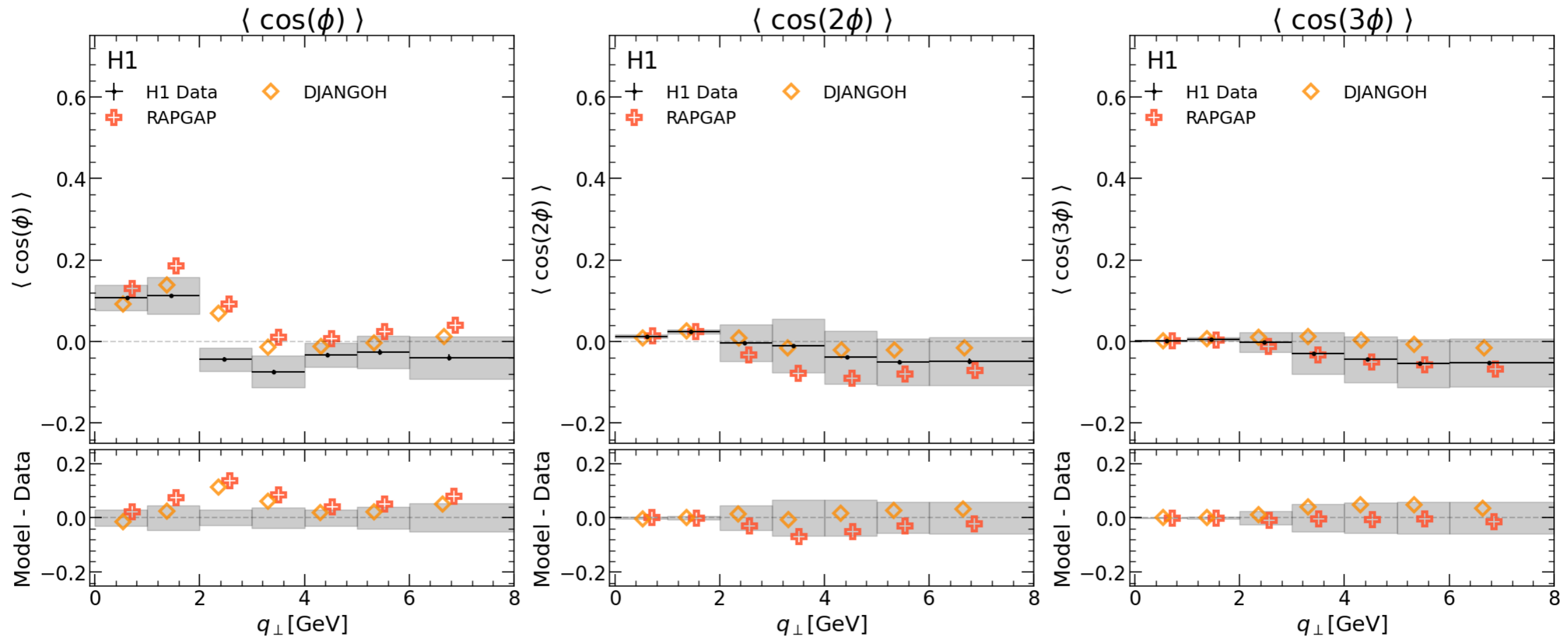
- Q^2 distribution simultaneously unfolds, displaying the energy scale dependance of observables, and yielding 30 unfolded distributions!

Moments of jet substructure



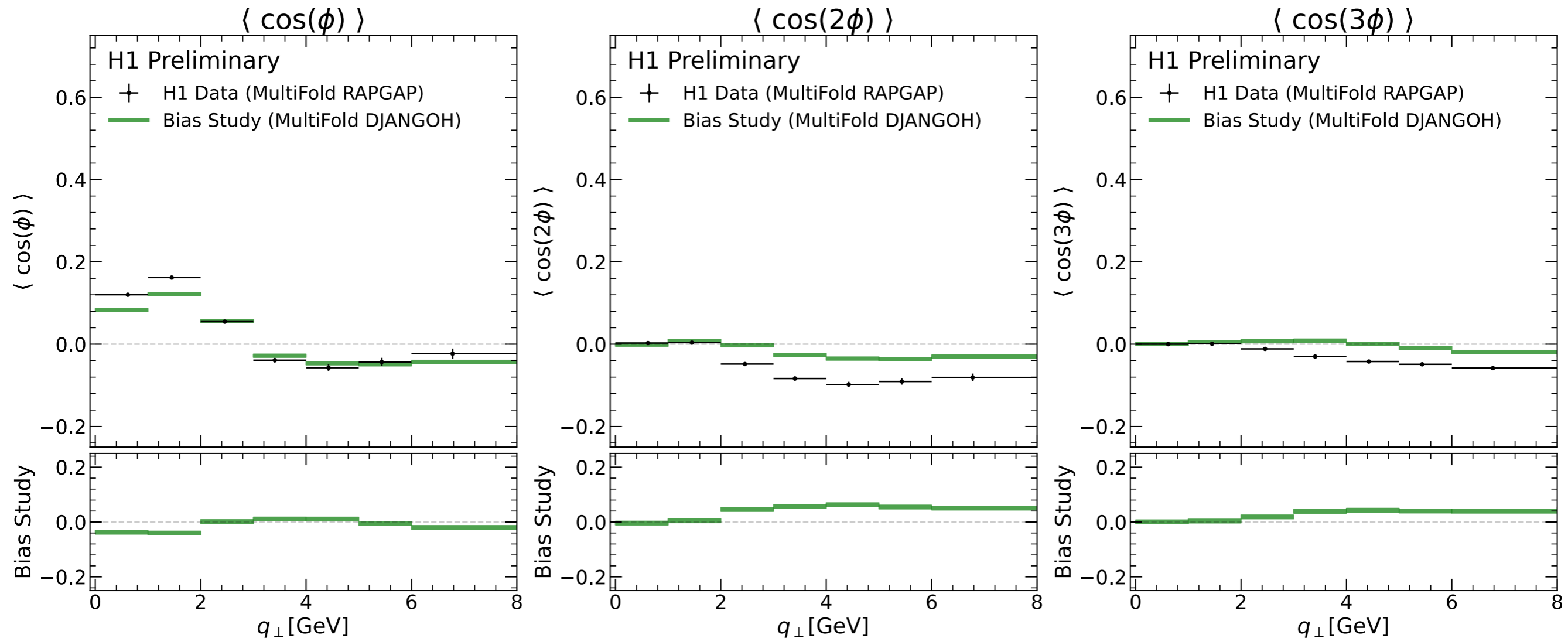
- Mean value of distributions also unfolded for free!
- Better agreement w/ generators at higher Q^2

H1 Unfolded Data + MC



- Leading moment is $\langle \cos(\phi) \rangle$, expected in lepton-jet events
- All harmonics approach 0.0 at higher q_{\perp} , may compromise $P_{\perp} \gg q_{\perp}$
- Rapgap and Django, tuned to HERA II, good agreement
- Breakdown of systemics next slide

Investigation of Model Bias vs. q_{\perp} [GeV]



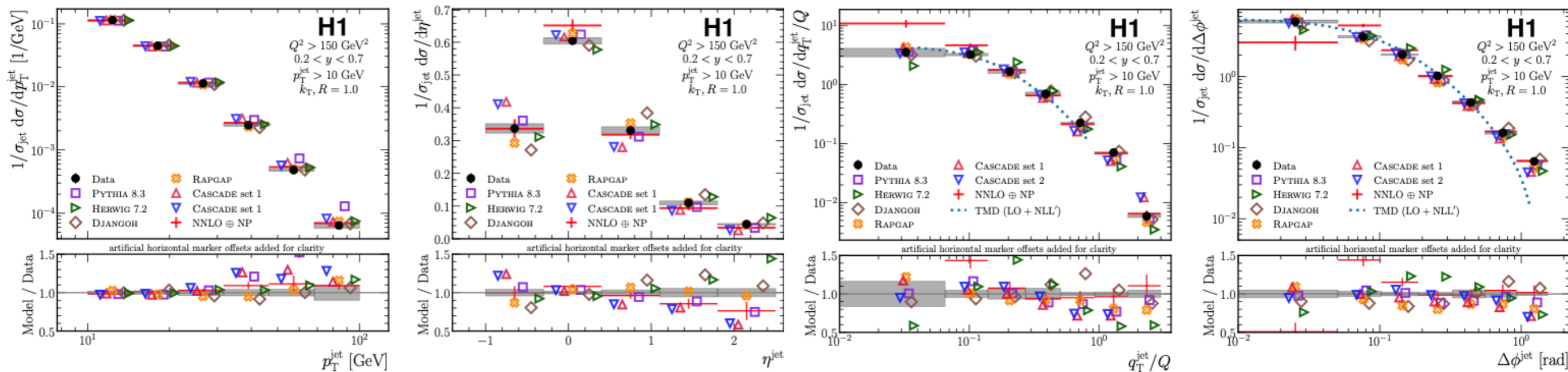
- Leading uncertainty is model bias in the unfolding for $\cos(2\phi)$ and $\cos(3\phi)$
- Difference in the result when unfolding using RAPGAP and DJANGO
- Reporting Abs. Errors; central values are very close to 0.0
- The Total Uncertainty is quite stable between harmonics

Systematic Uncertainties

- Model Dependence:
 - The bias of the unfolding procedure is determined by taking the difference in the result when unfolding using RAPGAP and DJANGO
 - The two generators have different underlying physics, thus providing a realistic evaluation of the procedure bias
- QED Radiation Corrections
 - Difference of correction between RAPGAP and DJANGO
 - Take RAPGAP with and without QED corrections
 - Take DJANGO with and without QED corrections
- Systematic uncertainties are determined by varying an aspect of the simulation and repeating the unfolding
 - These values detail the magnitude of variation:
 - HFS-object energy scale: $\pm 1 \%$
 - HFS-object azimuthal angle: ± 20 mrad
 - Scattered lepton azimuthal: ± 1 mrad
 - Scattered lepton energy: $\pm 0.5 - 1.0 \%$

Further Background

- Machine learning (OmniFold) is used to perform an 8-dimensional, unbinned unfolding. Present four, binned results:
- Use the 8-dimensional result to explore the Q^2 dependence and any other observables that can be computed from the electron-jet kinematics



**Extracted from the same phase-space as Yao's analysis,
but reporting a different observable**

OmniFold

$$1. \quad \omega_n(m) = \nu_{n-1}^{\text{push}}(m) L[(1, \text{Data}), (\nu_{n-1}^{\text{push}}, \text{Sim.})](m)$$

$$\omega_n^{\text{pull}}(t) = \omega_n(m)$$

- Detector level simulation is weighted to match the data
- $L[(1, \text{Data}), (\nu_{n-1}^{\text{push}}, \text{Sim.})](m)$ approximated by classifier trained to distinguish the *Data* and *Sim.*

$$2. \quad \nu_n(t) = \nu_0(t) L[(\omega_n^{\text{pull}}, \text{Gen.}), (\nu_0, \text{Gen.})](t)$$

- Transform weights to a proper function of the generated events to create a new simulation
- $L[(\omega_n^{\text{pull}}, \text{Gen.}), (\nu_{n-1}, \text{Gen.})](t)$ approximated by classifier trained to distinguish Gen. with *pulled* weights from Gen. using $\text{weights}_{\text{old}} / \text{weights}_{\text{new}}$

Each iteration of step 2 learns the correction from the original ν_0 weights

Advantage: Easier implementation, no need to store previous ν_n model

Disadvantage: Learning correction from ν_0 is more computationally expensive

IBU Generalization

$$\begin{aligned} t_j^{(n)} &= \sum_i \Pr_{n-1}(\text{truth is } j | \text{measure } i) \Pr(\text{measure } i) \\ &= \sum_i \frac{R_{ij} t_j^{(n-1)}}{\sum_k R_{ik} t_k^{(n-1)}} \times m_i, \end{aligned}$$

$$L[(w, X), (w', X')](x) = \frac{P_{(w, X)}(x)}{P_{(w', X')}(x)},$$

Differential Cross Section

- Back-to-back electron-jet production from ep collision,

$$e(l) + p(P) \rightarrow e(l') + J_q(p_J) + X$$

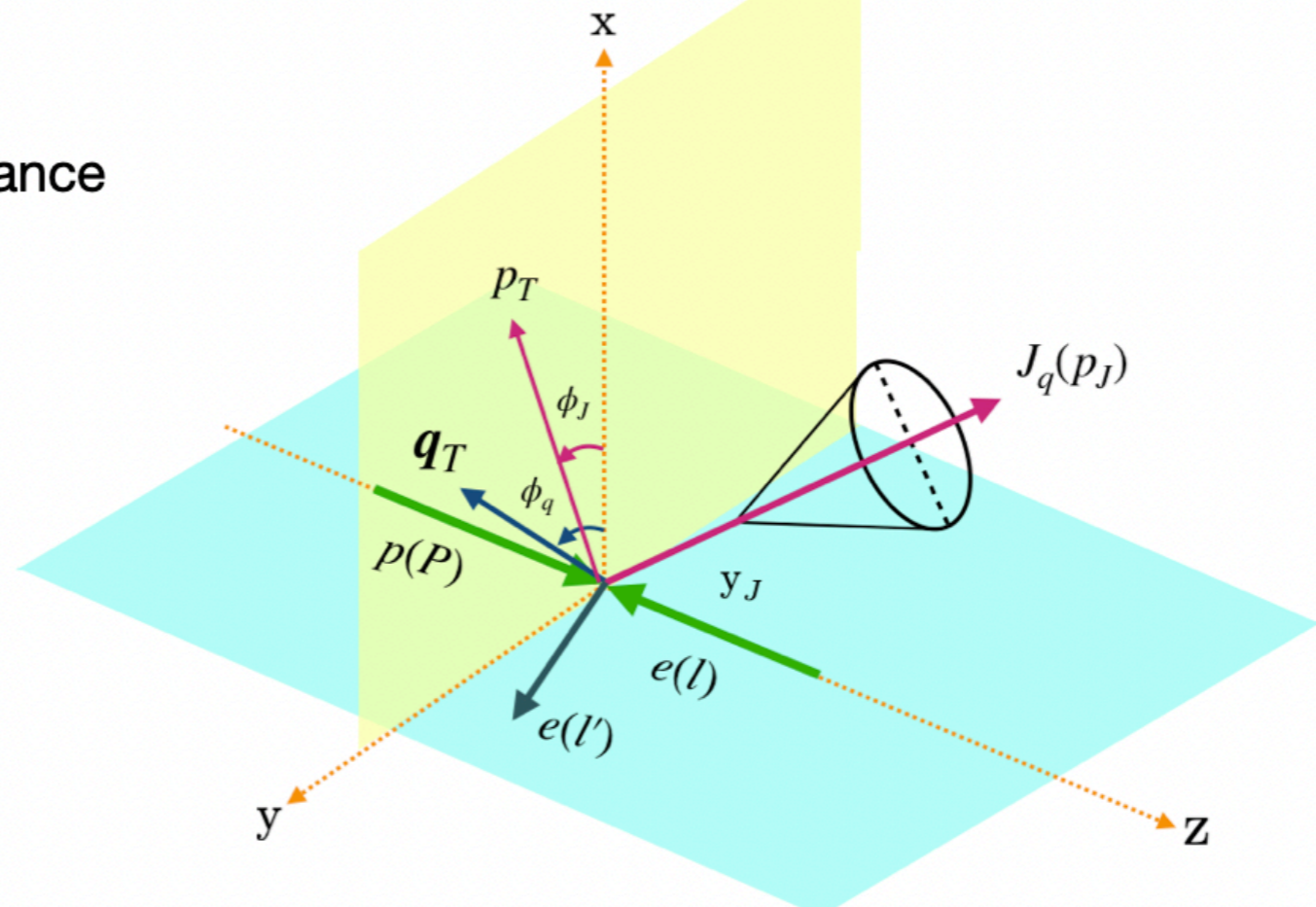
$$\frac{d\sigma}{d^2\mathbf{p}_T dy_J d\phi_J d^2\mathbf{q}_T} = \frac{d\sigma}{2\pi d^2\mathbf{p}_T dy_J q_T dq_T} \left[1 + 2 \sum_{n=1}^{\infty} v_n(p_T, y_T) \cos(n(\phi_q - \phi_J)) \right]$$

q_T : transverse momentum imbalance

$$\mathbf{q}_T = \mathbf{l}'_T + \mathbf{p}_{JT}$$

p_T : jet transverse momentum

y_J : jet rapidity



Note: slightly different angle definition, but background still applies]