Semi-inclusive DIS kinematic reconstruction with ML

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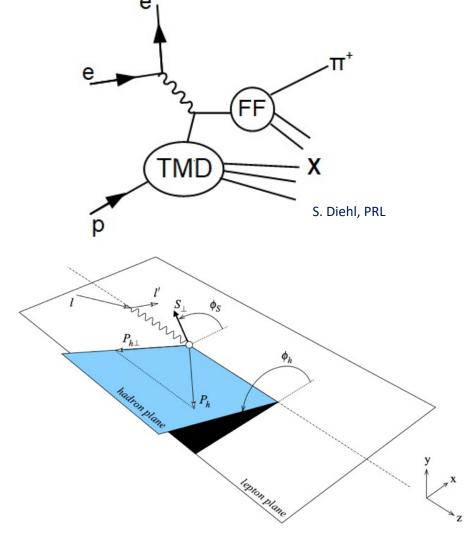






Semi-inclusive deep inelastic scattering at the EIC

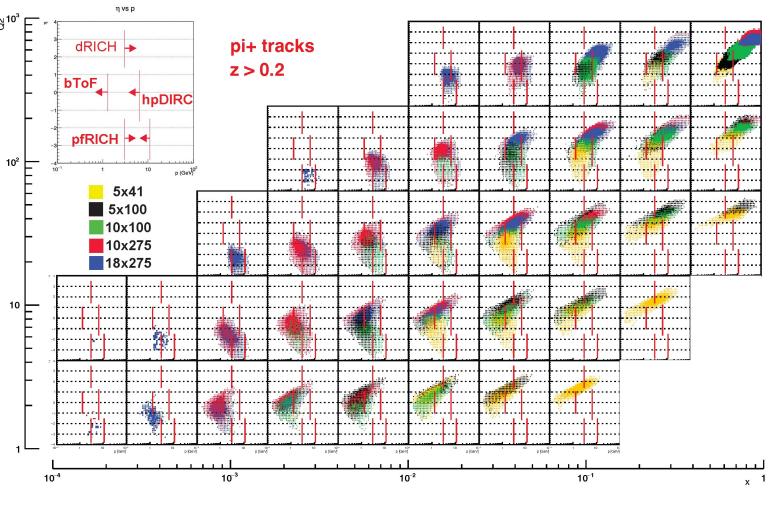
- SIDIS: virtual photon exchanged with parton, measure scattered lepton and single/di-hadrons
- SIDIS cross-section gives access to parton distribution functions and fragmentation functions
 - Extra degree of freedom from hadron vital for studying TMD-PDFs, TMD-FFs
 - Azimuthal angle and transverse momentum defined around virtual photon axis in target COM frame
 - Cross-section a function of $(x, y, z, p_T, \phi_h, \phi_s)$
- Many SIDIS observable projections made for ATHENA and ECCE proposals:
 - A_{LL} with kaons, gluon saturation with dihadrons, Sivers





SIDIS observables and coverage at the EIC

- Broad kinematics and PID coverage available at EIC/ATHENA
 - large lever arm for SIDIS multiplicities and asymmetries
- In addition to inclusive DIS reconstruction, rely on wide PID coverage (flavor separation), precise hadron tracking, virtual photon reconstruction



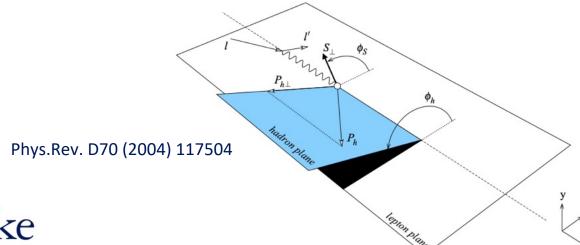


SIDIS kinematic reconstruction

• SIDIS variables: reliant on reconstruction of virtual photon four-momentum, typically determined using

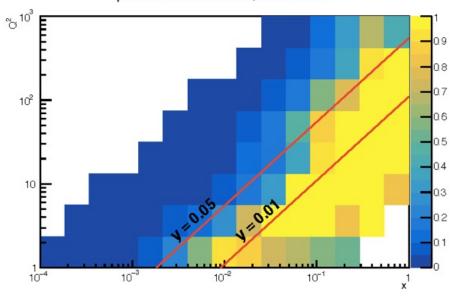
$$q = l - l'$$

- Reliable for larger y, but begins to fail for y < ~0.05
 Low-y: region of interest for TMDs, and important for evolution studies
- To utilize full EIC kinematic reach for SIDIS studies, need improved methods to determine SIDIS variables
- CC would require first method without electron



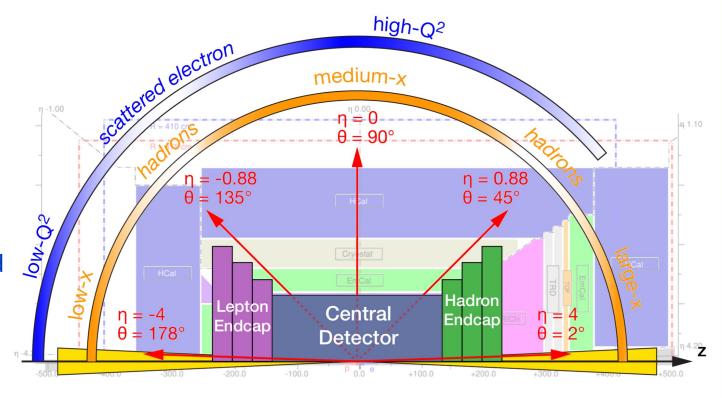
ATHENA full simulation:

pT mean relative error, ele. method



Reconstruction with hadronic final state

- Through conservation of momentum and energy, hadronic final state (HFS) should also contain enough information to constrain q
- HFS methods developed at HERA for inclusive DIS variables for regions in which electron method less reliable
- Methods utilizing hadronic final state should be more robust with respect to first-order radiative corrections
 - Impact of radiative effects expected to be large for SIDIS observables at EIC





Reconstruction with hadronic final state

- Method used in EIC YR and ATHENA proposal to reconstruct virtual photon using hadronic final state (HFS)
 - x and y components summed HFS momentum
 - z and t components solved for algebraically using

$$y = \frac{p \cdot q}{p \cdot l} \qquad Q^2 = -q^2$$

and DIS variables from any DIS reconstruction method

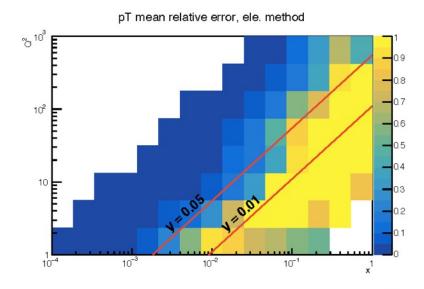
- $\begin{array}{lll} i) & Leptonic \ variables & q \equiv q_l = k_2 k_1, & y_l = p_1.(k_1 k_2)/p_1.k_1 \\ ii) & Hadronic \ variables \ [81] & q \equiv q_h = p_2 p_1, & y_l = p_1.(p_2 p_1)/p_1.k_1 \\ iii) & Jacquet\text{-Blondel variables} \ [82] & Q_{JB}^2 = (\vec{p}_{2,\perp})^2/(1-y_{JB}), & y_{JB} = \Sigma/(2E(k_1)) \\ & \Sigma = \sum_h (E_h p_{h,z}) & \text{Prog. Part. Nucl. Phys. 2013, Blümlein} \\ iv) & Mixed \ variables \ [81] & q = q_l, y_m = y_{JB} \\ v) & Double \ angle \ method \ [83] & Q_{DA}^2 = \frac{4E(k_2)^2 \cos^2(\theta(k_2)/2)}{\sin^2(\theta(k_2)/2) + \sin(\theta(k_2)/2) \cos(\theta(k_2)/2) \tan(\theta(p_2)/2)}, \\ & y_{DA} = 1 \frac{\sin(\theta(k_2)/2) + \cos(\theta(k_2)/2) \tan(\theta(p_2)/2)}{\sin(\theta(k_2)/2) + \cos(\theta(k_2)/2) \tan(\theta(p_2)/2)}, \end{array}$
- + solution in quadratic equation found to always be closer to MC truth
- Resolution improved if this is carried out in head-on frame, then transformed to lab frame

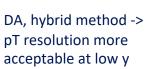


Large crossing angle needed for EIC

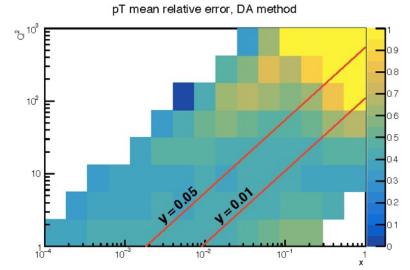
ATHENA full simulation SIDIS resolution, p_T

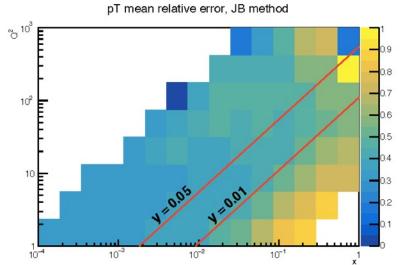
Transverse momentum (w.r.t. q), 10x275, pi+, z > 0.2:







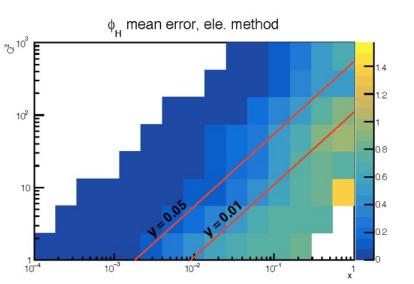


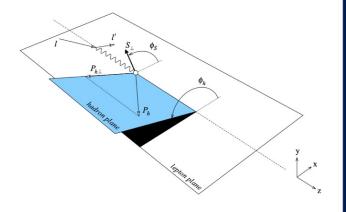


<- JB, HFS only Potential for CC

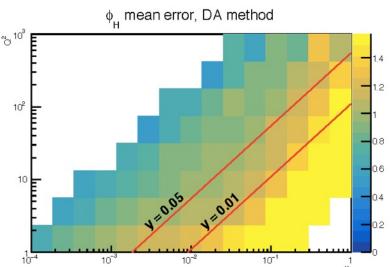
ATHENA full simulation SIDIS resolution, ϕ_h

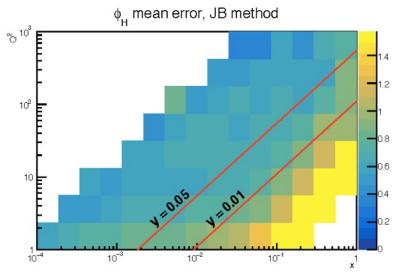
ATHENA full simulation, 10x275, pi+, z > 0.2





Angular resolution still poor at low-y with all methods







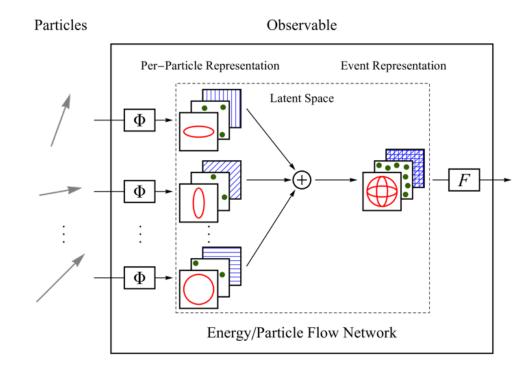
Machine learning reconstruction

- Based on hybrid HFS-electron SIDIS reconstruction, using ML to combine information from both to reconstruct q
 - Potential to correct overall HFS and electron momentum and to more reliably reconstruct z and t components than exact formula
 - ML models used for DIS reconstruction have been shown to be able to naturally account for radiative effects
 - (arXiv:2108.11638 Diefenthaler, Farhat, Verbytskyi, Xu, as well as NIM-A 1025 (2022) 166164, Arratia, Britzger, Long, Nachman)
- Currently utilizing graph-like neural network architectures designed for jet reconstruction



Particle flow networks

- Particle flow networks (PFN) developed by Komiske et al., (JHEP 01 (2019) 121, Komiske, Metodiev, Thaler)
 - Accepts unordered set of particles
 - Particles -> input to layers Φ
 - Summed over to created latent space of ℓ variables
 - Global features of event concatenated with latent space variables
 - Latent space variables and global features fed to layers F, produce final output
- Designed to be general purpose, universal approximator on sets of particles



JHEP 01 (2019) 121



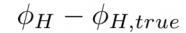
ML SIDIS model and training

- Model combining electron and HFS:
 - Particle features for PFN: momentum, energy, η , ϕ in lab frame
 - Event-wide features: electron four momentum, DIS variables from JB, DA, electron methods
 - DIS variables will eventually be replaced with final reconstructed Q2 and x likely using another ML method, but in this study statistics for training were limited
 - Target: MC virtual photon four-momentum in lab frame
- Training sample: ATHENA full simulation
 - Version of dd4hep ATHENA full sim. used for detector proposal
 - Still some features missing, e.g. proper scattered electron ID
 - HFS at the level of reconstructed particles
 - 10 GeV electron beam, 275 GeV proton beam, crossing angle -25 mrad
 - Trained on 3 million events with $Q^2 > 1$ GeV², 2 million with $Q^2 > 10$ GeV²
 - 1 million Q² > 1 GeV² events for validation

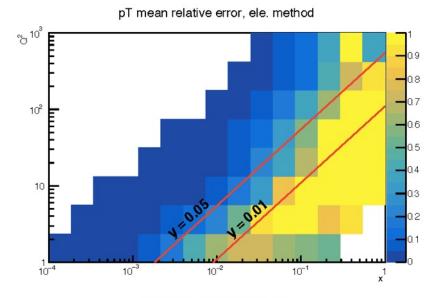


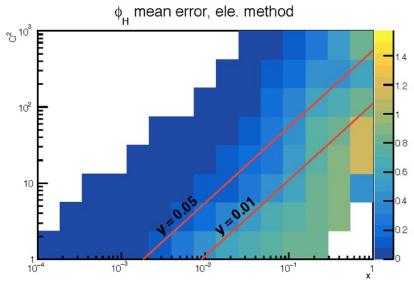
ATHENA full simulation, 10x275, pi+, z > 0.2

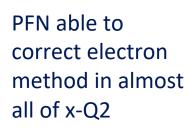
$$\frac{p_T - p_{T,true}}{p_{T,true}}$$



Electron method

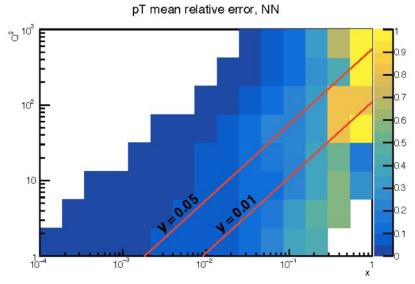


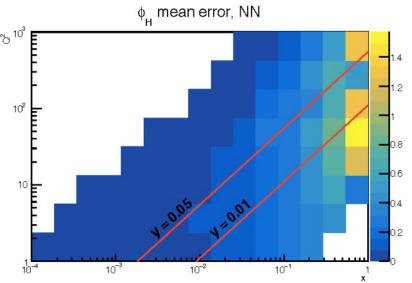




Neural network





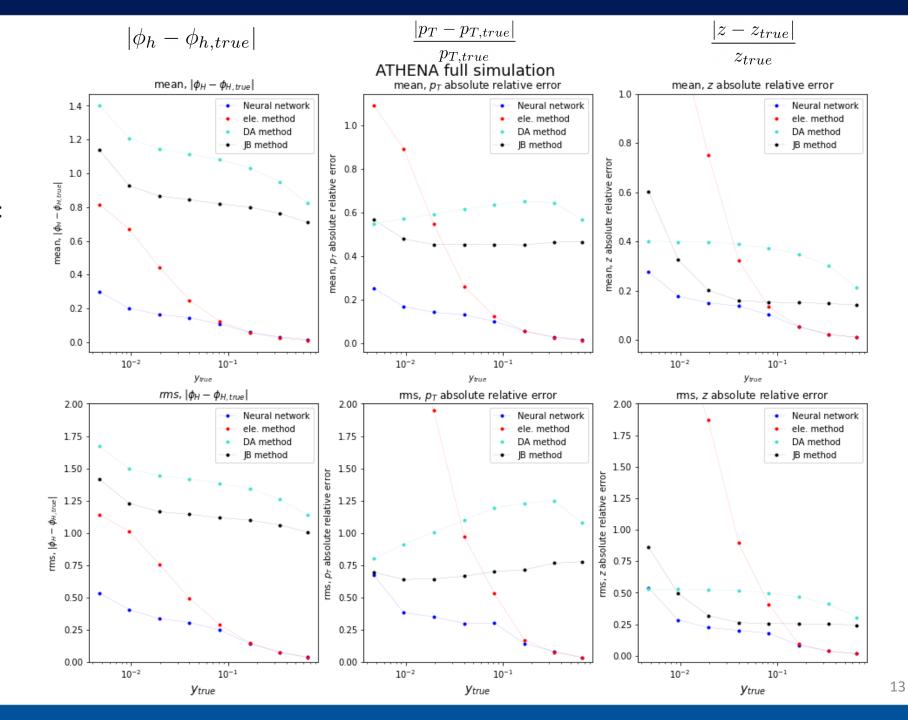


ATHENA full simulation, 10x275, pi+, z > 0.2 Reconstruction as a function of y

Mean:

- Comparison with other HFS/hybrid methods vs
 - **Y**true
- NN clearly best performance for low y, and at least equaling electron method for large y

RMS:



Duke

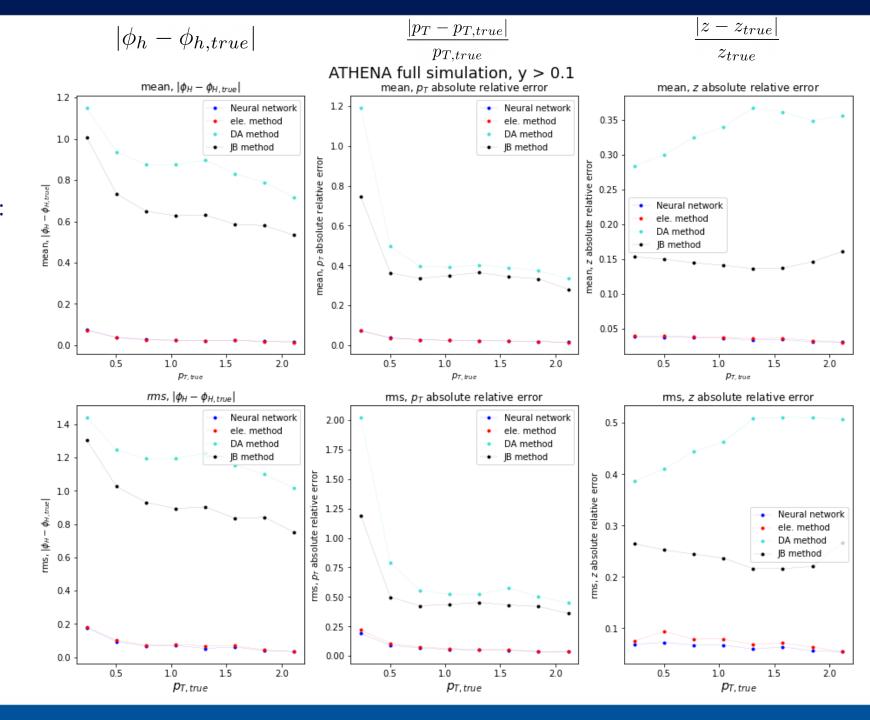
ATHENA full simulation, 10x275, pi+, z > 0.2 Reconstruction as a function of pT, y > 0.1

Mean:

- Comparison with other HFS methods vs true pT
- Machine learning method equals/slightly outperforms electron method for large y (expected)

RMS:





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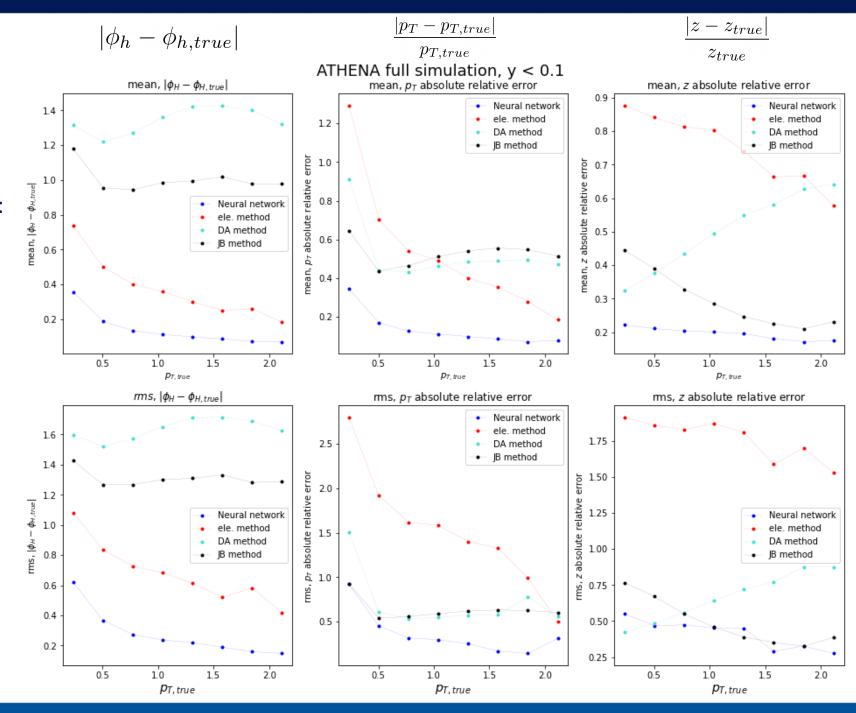
ATHENA full simulation, 10x275, pi+, z > 0.2 Reconstruction as a function of pT, y < 0.1

Mean:

- Comparison with ele., other HFS methods vs true pT
- At low-y, network is able to reliably reconstruct kinematics even at very low pT (w.r.t virtual photon)
- Mean pT ~ 0.5 GeV

RMS:

Duke



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Summary

- Projections for the ATHENA and ECCE detectors demonstrate the exciting capabilities and kinematic coverage of the EIC for SIDIS measurements
- The electron method fails for y < 0.05, but can be improved using the hadronic final state and DIS variables to reconstruct virtual photon axis
- We demonstrate a machine learning approach combining the hadronic final state and scattered electron which surpasses existing methods for all of x-Q2 and p_T
- Next steps in reconstruction:
 - Currently working on replacing the particle flow network with an architecture which can learn correlations between HFS particles (such as a GNN), as well as exploring other deep learning approaches
 - Method will need to be tested with better implementation of radiative effects

Research supported by:



