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Centro Nazionale di Ricerca in HPC,
Big Data and Quantum Computing



Istituto Nazionale di Fisica Nucleare

Measurement of inclusive DDIS in EIC and determination of DPDFs with EIC Pseudodata

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Jan. 23, 2025

ePIC collaboration meeting

Outline:

- Motivation
- What has been so far
- What is the plan for future



Theory: Opportunities for first EIC physics

P. Zurita



ePIC collaboration meeting
Jan. 20-24, 2025



Early Science - Inclusive diffraction

Inclusive Diffraction with Sartre Current Status

Current version on SVN:

Can generate events with both $q\bar{q}$ and $q\bar{q}g$ final states in ep and eA

To Do (short term):

Implement saturation effects in final state
Create full tables for several initial state species

Thorough testing

To Do (intermediate term):

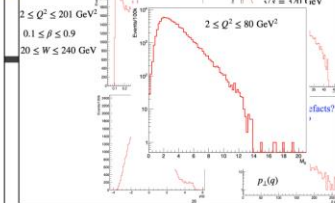
Implement t -dependence

To Do (long term):

Incoherent Diffraction?



Event Generation Sartre



Inclusive DDIS in eP/eA process:

$$Q^2 = -q^2 = (l' - l)^2 \quad y = \frac{P \cdot q}{P \cdot l} \quad x = \frac{Q^2}{2P \cdot q}$$

$$x_{\mathbb{P}} = \frac{2(P - P') \cdot q}{P \cdot q} \quad \beta = \frac{Q^2}{2(P - P') \cdot q} = \frac{x}{x_{\mathbb{P}}}$$

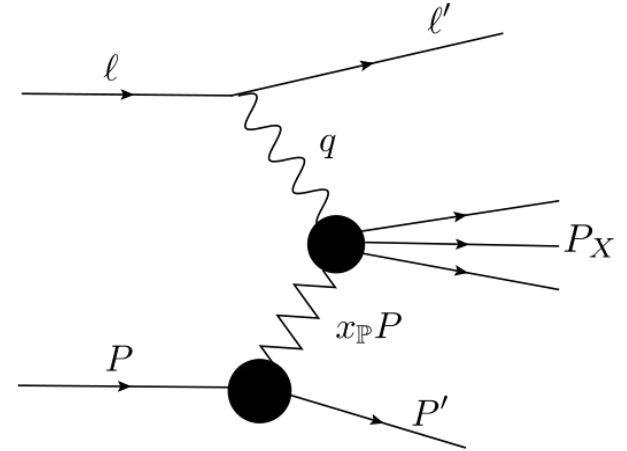
Triple differential Cross section

$$\frac{d\sigma^{ep \rightarrow epX}}{d\beta dQ^2 dx_{\mathbb{P}}} = \frac{2\pi\alpha^2}{\beta Q^4} [1 + (1 - y)^2] \sigma_r^{D3}(\beta, Q^2; x_{\mathbb{P}})$$

$$\sigma_r^{D3}(\beta, Q^2; x_{\mathbb{P}}) = F_2^{D3}(\beta, Q^2; x_{\mathbb{P}}) - \frac{y^2}{1 + (1 - y)^2} F_L^{D3}(\beta, Q^2; x_{\mathbb{P}})$$

Using the factorization theorem on Fracture functions

$$F_k^{D3}(\beta, Q^2; x_{\mathbb{P}}) = \sum_i \mathcal{F}_i^{D3}(\beta, Q^2; x_{\mathbb{P}}) \otimes C_{ki}(\beta, Q^2, \alpha_S)$$



Fracture functions: An Improved description of inclusive hard processes in QCD Phys. Lett. B 323, 201 (1994)

Monte Carlo Event Generators:

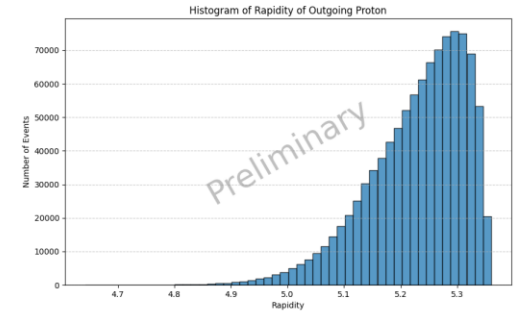
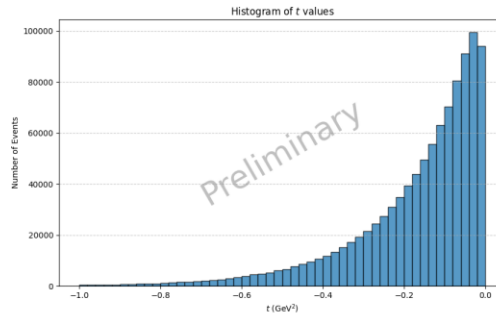
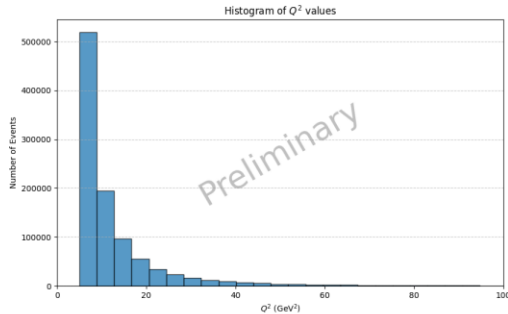
- We are going to use RAPGAP for event generation
- Event generation based on $10 \text{ fb}^{-1} = 10^7 \text{ nb}^{-1}$ luminosity and $\sigma = 4.22 \text{ nb}$:
- - number of events:
 $4.22 \text{ (nb)} * 10^7 \text{ (nb}^{-1}) \approx 4.2 \times 10^7$ events
- 42 million events as mock data.
- 42 million events as MC data, totaling 84 million events.
- Proton momentum : 100 GeV
- Electron momentum : 10 GeV
- SARTRE for ep and eA inclusive, beta soon(?)
Tobias Toll talk @ Exclusive, Diff. and tagging WG
- Then we can have a nice comparison between these two

<https://indico.bnl.gov/event/25025/contributions/100107/attachments/59181/101676/SartreInclusiveDiffraction.pdf>



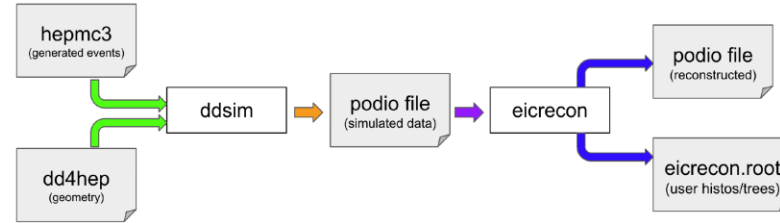
Control distributions:

- **Q² Distribution:**
- **t Distribution:**
- **Rapidity of Outgoing Proton:**
- **Overall:**
 - Dominant forward-scattering component with minimal momentum transfer, characteristic of diffractive and elastic processes



Update on event generation:

- The workflow of detector simulation and reconstruction needs HepMC3 format
- We found out that the version 3.310 of RAPGAP with HepMC support has a bug, in some events there are more than 2 particles designated as beam!
- We are in contact with Hannes Jung about this



```
E 2 11 28
U GEV MM
W 1.000000000000000000000000e+00
A 0 GenCrossSection 4.22617769e+03 1.23955151e+01 -1 -1
A 0 GenPdfInfo 11 2212 3.07580054e-01 5.40348813e-02 6.64962981e+01 -3.07580042e+00 5.40348816e+00 252 10041
A 0 alphaQCD -1
A 0 alphaQED -1
A 0 event_scale 66.4963
A 0 mpi -1
A 0 random_states 0
A 0 signal_process_id 0
A 0 signal_process_vertex 0
P 1 0 11 1.7763568394002505e-15 0.000000000000000000e+00 -9.9999999999999999e+00 1.0000000013005049e+01 5.1000000000000004e-04 4
P 2 1 23 2.4461285218615494e+00 -6.329263442228229e+00 -4.7382444723232906e+00 1.4134295707586053e+00 -8.1545262274720809e+00 3
P 3 1 11 -2.4461285218615494e+00 6.3292634422282264e+00 -5.2617555276767298e+00 8.5865704422463729e+00 5.1000000000000004e-04 1
P 4 0 2212 -1.7763568394002505e-15 3.5527136788005009e-15 1.0000000000000000e+02 1.0000440165609162e+02 9.3827000000000005e-01 4
P 5 4 990 1.6697379176052896e-01 -2.8478672883558431e-03 1.4687212367650162e+01 1.4686291195575638e+01 -2.3440650310167599e-01 3
P 6 4 21 2.7739710433714315e-01 1.0555810530099841e-01 1.0997649207188017e+01 1.1001052652471973e+01 -0.0000000000000000e+00 3
P 7 4 2212 -1.6697267947604644e-01 2.84777991491580045e-03 8.5312233960207607e+01 8.5317556820736741e+01 9.3827000000000005e-01 1
P 8 4 21 -4.5373307959223674e-02 -7.2619797935561059e-02 3.6628929682694622e+00 3.7301631669989908e+00 6.9999998807907104e-01 2
P 9 2 1 3.5291873685974604e+00 -5.2447542032833976e+00 7.5698783562734562e-01 6.3753071077930450e+00 3.3000000000000002e-01 3
P 10 9 1 3.5291873685974604e+00 -5.2447542032833976e+00 7.5698783562734562e-01 6.3753071077930450e+00 3.3000000000000002e-01 2
P 11 0 21 2.6442149021873207e-01 -5.6189047227182543e-02 2.4018956585142837e+00 2.5163820598665572e+00 6.9999998807907104e-01 4
P 12 0 -1 -1.1351332572460775e+00 -9.5854821236685406e-01 3.1271913558982938e+00 3.4778683073892824e+00 3.3000000000000002e-01 4
```

Afterburner will complain here

Conclusions and Future Directions

- We are working with RAPGAP v3.310 and waiting for SARTRE to be ready for inclusive diffraction.
- We believe that using RAPGAP/SARTRE inclusive events provide a good comparison that gives us insight into early science study.
- We think that a neural network parameterization, combined with new pseudodata from ePIC, will improve the precision and reliability of DPDF extraction beyond current limits.
- Progress on fitting code and ROOT analysis scripts is ongoing.



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Thank
You!

Backup slides

Neural Network Approach to Parameterization

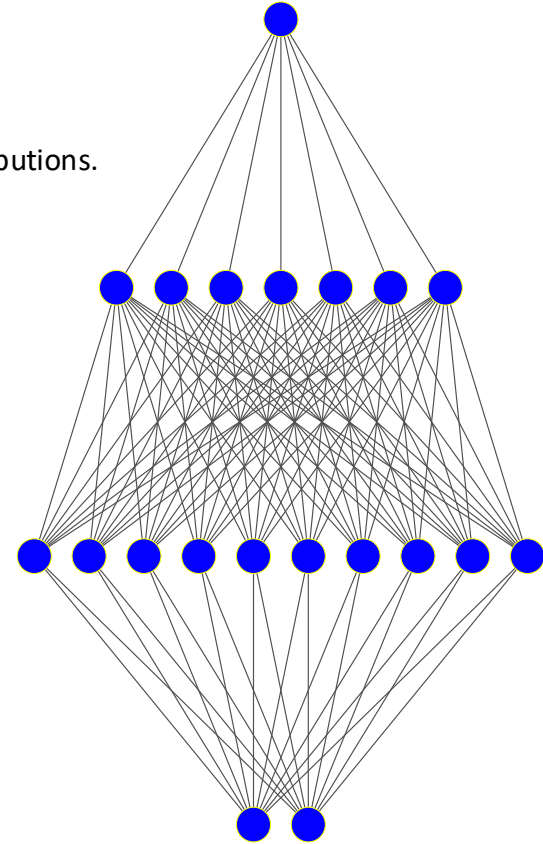
- **A Feed-forward neural network** with two hidden layers used to parametrize diffractive distributions.
- Benefits of using NNs: flexibility, no need for strict theoretical assumptions.

$$\beta \mathcal{F}_q^D(\beta, Q_0^2; x_{\mathbb{P}}) = \mathcal{W}(x_{\mathbb{P}})(NN_1(\beta, Q_0^2) - NN_1(1, Q_0^2))^2$$

$$\beta \mathcal{F}_g^D(\beta, Q_0^2; x_{\mathbb{P}}) = \mathcal{W}(x_{\mathbb{P}})(NN_2(\beta, Q_0^2) - NN_2(1, Q_0^2))^2$$

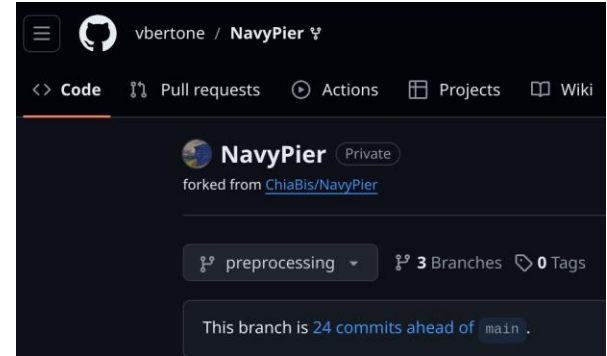
$$Q_0^2 = 2 \text{ GeV}^2$$

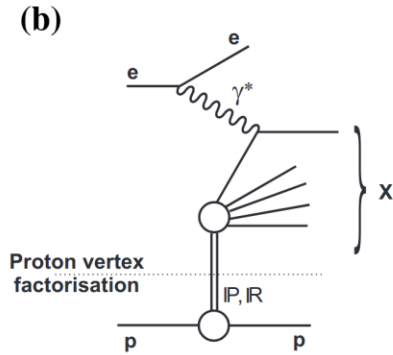
$$\mathcal{W}(x_{\mathbb{P}}) = x_{\mathbb{P}}^{w_1}(1 - x_{\mathbb{P}})^{w_2}(1 + w_3 x_{\mathbb{P}}^{w_4})$$



DPDF analysis: Computational Tools

- **Code Name:** NavyPier
- **Programming Language:** C++
- **Based On:** APFEL++
- Used for **DGLAP evolution** and **convolution integrals**.
- **Core Functionalities:**
 - **DGLAP Evolution:** Solves equations related to PDFs using APFEL++.
 - **Convolution Integrals:** Computes integrals using APFEL++.
 - **Parameterization:** Utilizes a feed forward NN library NNAD.
- **Optimization Tool:** Uses Google's **CERES Solver** to minimize the chi-squared function.
- **Uncertainty Estimation:** Implements the Monte Carlo replica method.





$$F_{2/L}^{D(4)}(\beta, Q^2; x_{IP}, t)$$

$$= \sum_i \int_{\beta}^1 \frac{dz}{z} C_{2/L,i} \left(\frac{\beta}{z} \right) f_i^D(z, Q^2; x_{IP}, t),$$

$$f_i^D(x, Q^2, x_{IP}, t) = f_{IP/p}(x_{IP}, t) \cdot f_i(\beta = x/x_{IP}, Q^2)$$

parameterized by NNs

$$f_{IP,IR}(x_{IP}, t) = A_{IP,IR} \frac{e^{B_{IP,IR} t}}{x_{IP}^{2\alpha_{IP,IR}(t)-1}},$$

Update on fitting code:

Standard Pomeron-Reggeon approach is added to the fitting code, (testing)

```

double xp = 0.003; // this is for substitution in A_P see hep-ex/0606004 eq. (14) and below
// Calculate the quantity
double denom_A_Pom = (-((std::pow(xp, 2 * alpha_val)) / std::exp(B_val)) + (std::exp((B_val * m_val * m_val * xp *
(std::pow(xp, (2 * m_val * m_val * xp * xp * alpha_val) / (-1 + xp))))));
double A_Pom = ((std::pow(xp, -2 + 2 * alpha_val)) * (B_val - 2 * alpha_val * std::log(xp))) / denom_A_Pom;

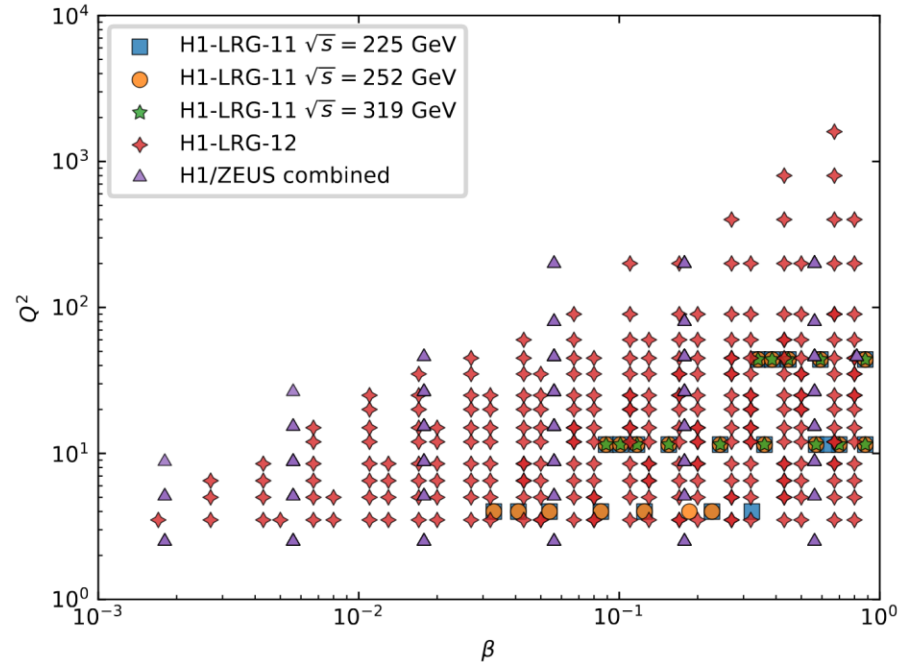
// we are going to multiply the flux factor by xPom, because the observable we are interested in is xPom*sigma_red
return std::vector<double>{
// Expression
xp_val * A_Pom * ( xp_pow_1_minus_2_alpha0 * (-xp_alpha_term / exp_B) + common_exp / xp_alpha_common) / denom
,

// Derivatives
// Derivative w.r.t. B

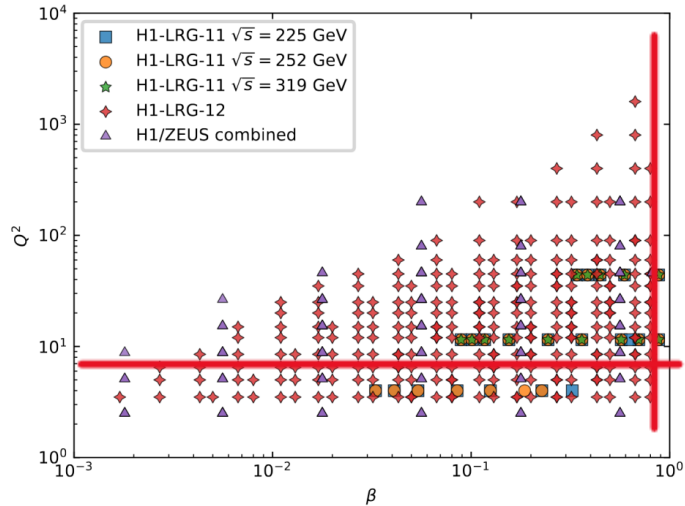
```

Diffractive DIS data sets from HERA

- F. D. Aaron et al. [H1 Collaboration],
Measurement of the diffractive longitudinal structure function FLD at HERA. Eur. Phys. J. C **72**, 1836 (2012).
- F. D. Aaron et al. [H1 Collaboration],
Measurement of the diffractive longitudinal structure function FLD at HERA. Eur. Phys. J. C **71**, 1836 (2011).
- F. D. Aaron et al. [H1 Collaboration],
Inclusive measurement of diffractive deep-inelastic scattering at HERA. Eur. Phys. J. C **72**, 2074 (2012).
- F. D. Aaron et al. [H1 and ZEUS Collaborations],
Combined inclusive diffractive cross sections measured with forward proton spectrometers in deep inelastic ep scattering at HERA. Eur. Phys. J. C **72**, 2175 (2012).



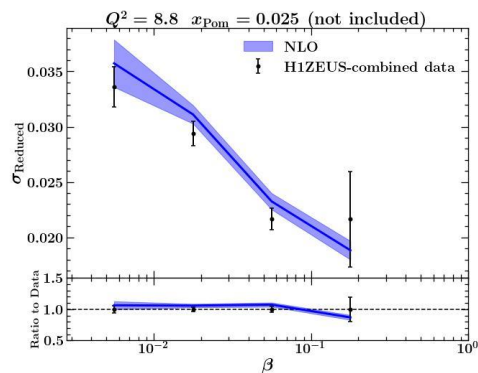
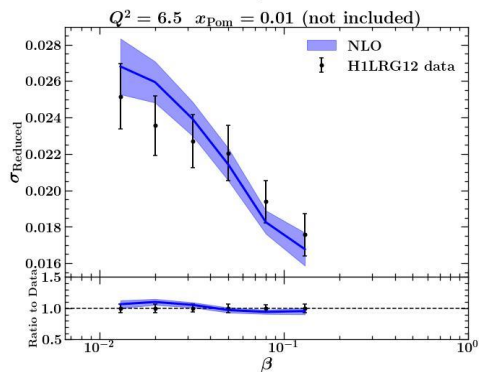
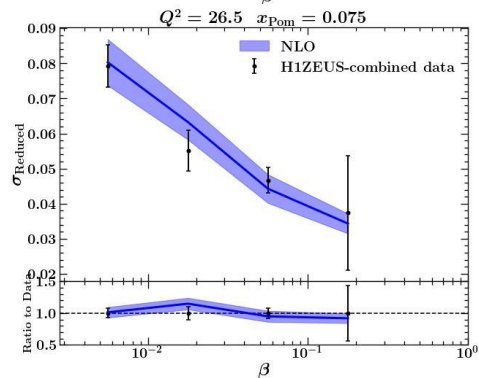
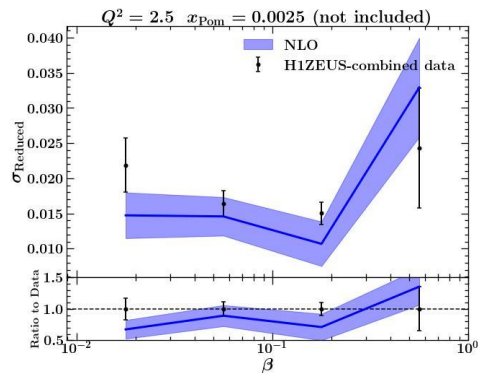
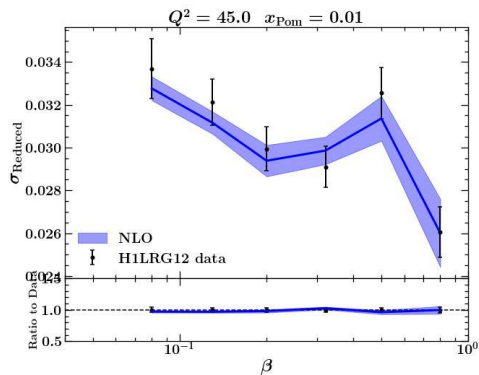
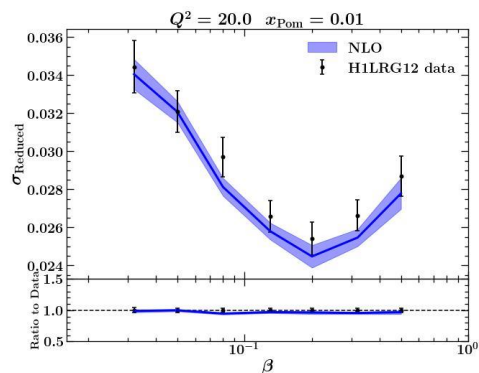
Training the Neural Network



- NN trained using **HERA reduced cross-section data**. NN adjusts weights iteratively, learning the correct mapping between input momentum fraction and DPDFs.
- **Optimization**: Chi-squared minimization between predicted and observed diffractive cross-sections.
- We use H1 Large rapidity gap data and H1/Zeus combined proton spectrometer data.
- Kinematical cuts:
 - $\beta \leq 0.80$
 - $Q^2 > 8.5$ GeV² for all the datasets
 - After the cuts we have 301 datapoints
 - Total chi-squared over number of datapoints: 0.88

Results

- Fair agreement between **data** and NN predictions.



Results

- NN results show a fair matching with previous DPDF determinations.
- **Monte Carlo method** applied for uncertainty estimation: generating pseudo-datasets (replicas) to calculate central values and uncertainties.

