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# Towards a Neural Network determination of nuclear PDFs

R. Abdul Khalek, J. Ethier, J. Rojo

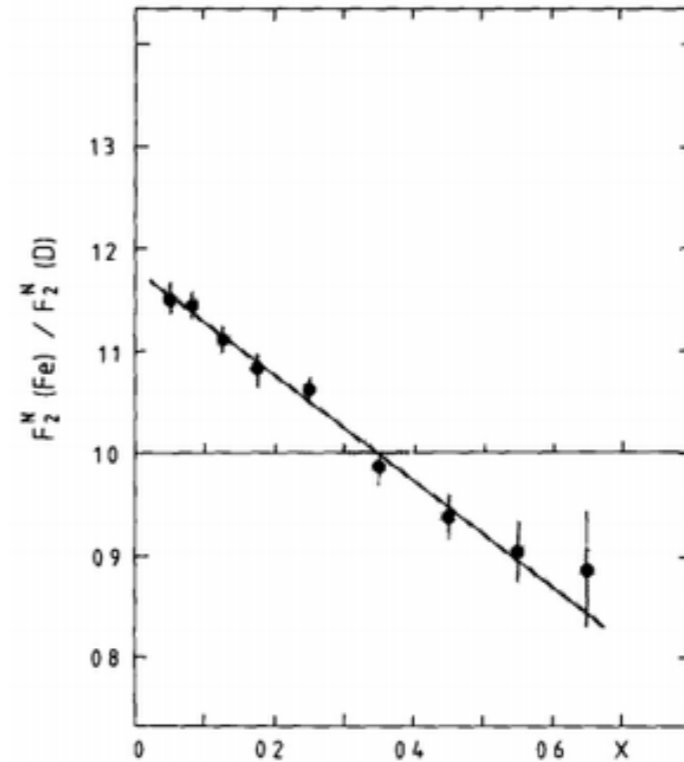
Departement of Physics and Astronomy, Vrije Universiteit Amsterdam  
Nikhef Theory Group

# HP18

Hard Probes 2018  
1 - 5 October

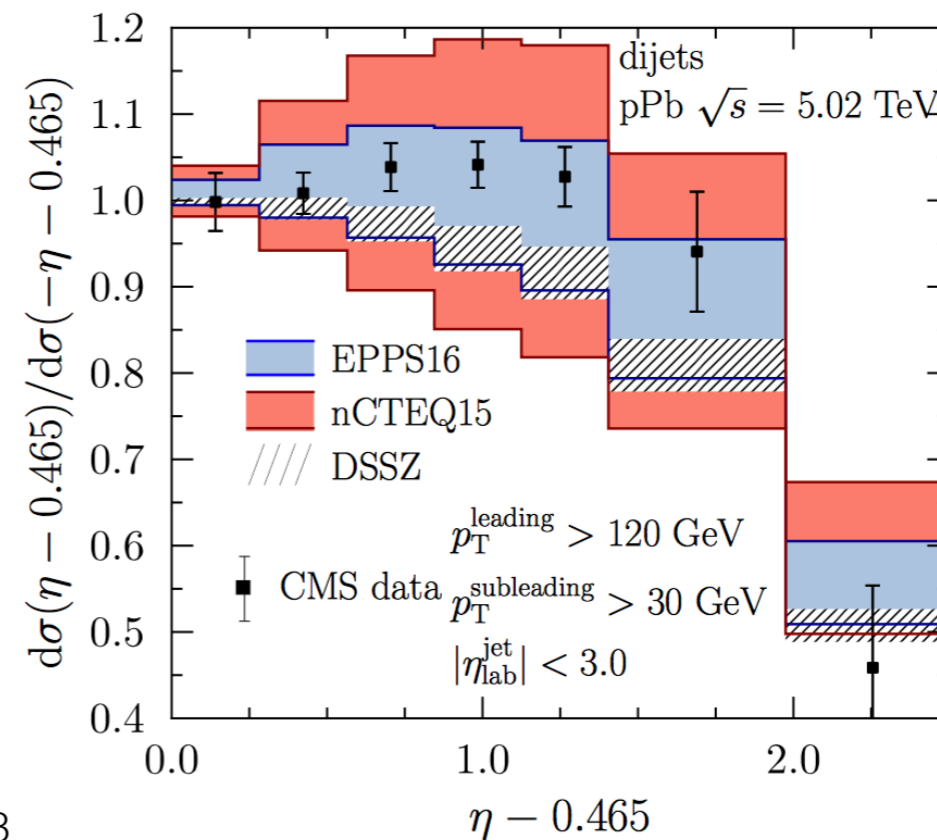
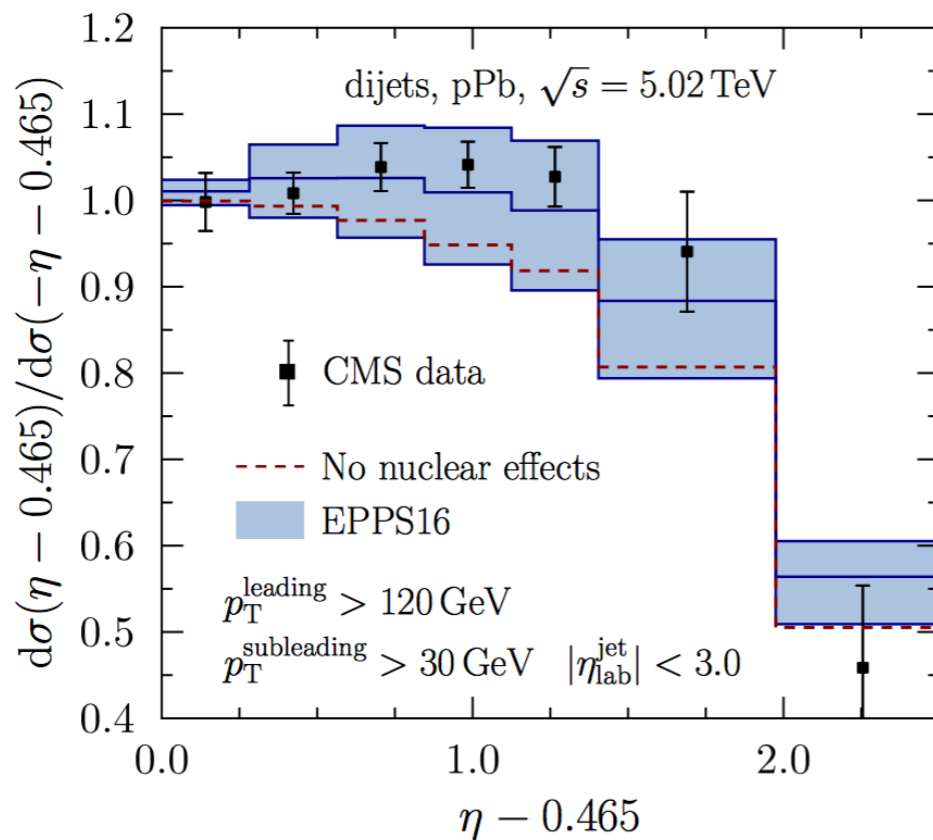
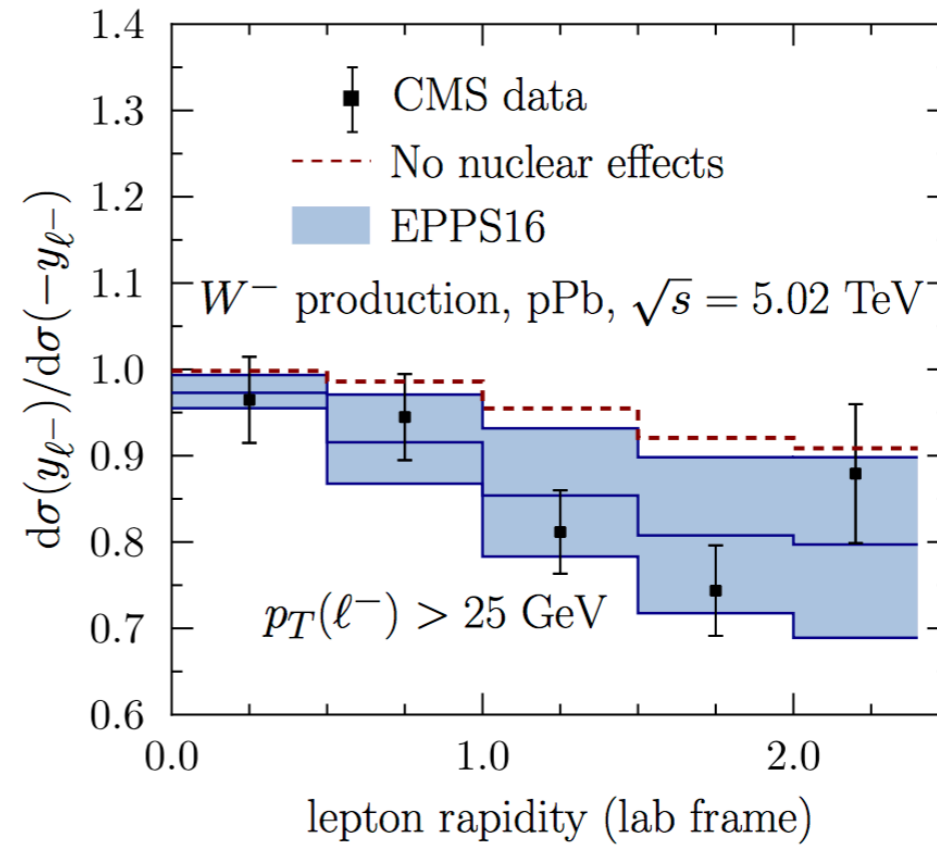
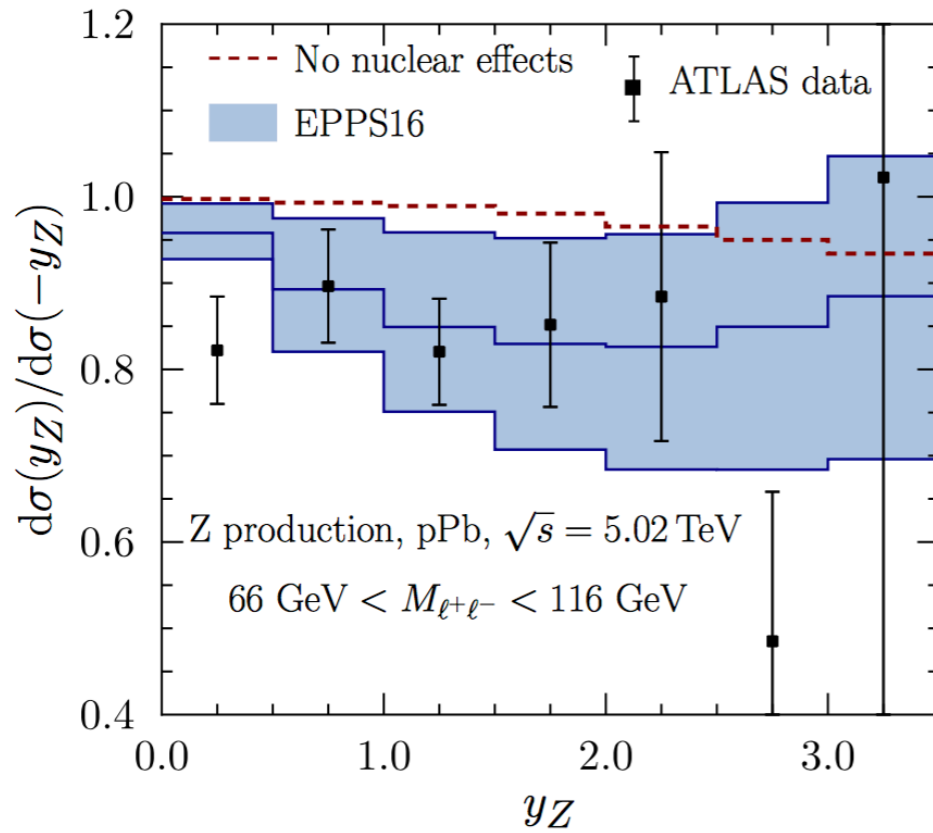
# Physics Motivation

- **Simulations** of relativistic nucleus-nucleus reactions [LHC, RHIC, etc.]
- Understanding the **EMC effect**.
- **Lead, Xenon PDF** for the LHC heavy-ion program.
- Combining data across different nuclear targets allows testing **different nuclear models**.
- Heavy nuclear targets in  $\nu$ -A DIS measurements provide important information on the quark **flavour separation** in proton PDF fits.



# nPDF in LHC pPb observables

[1612.05741]



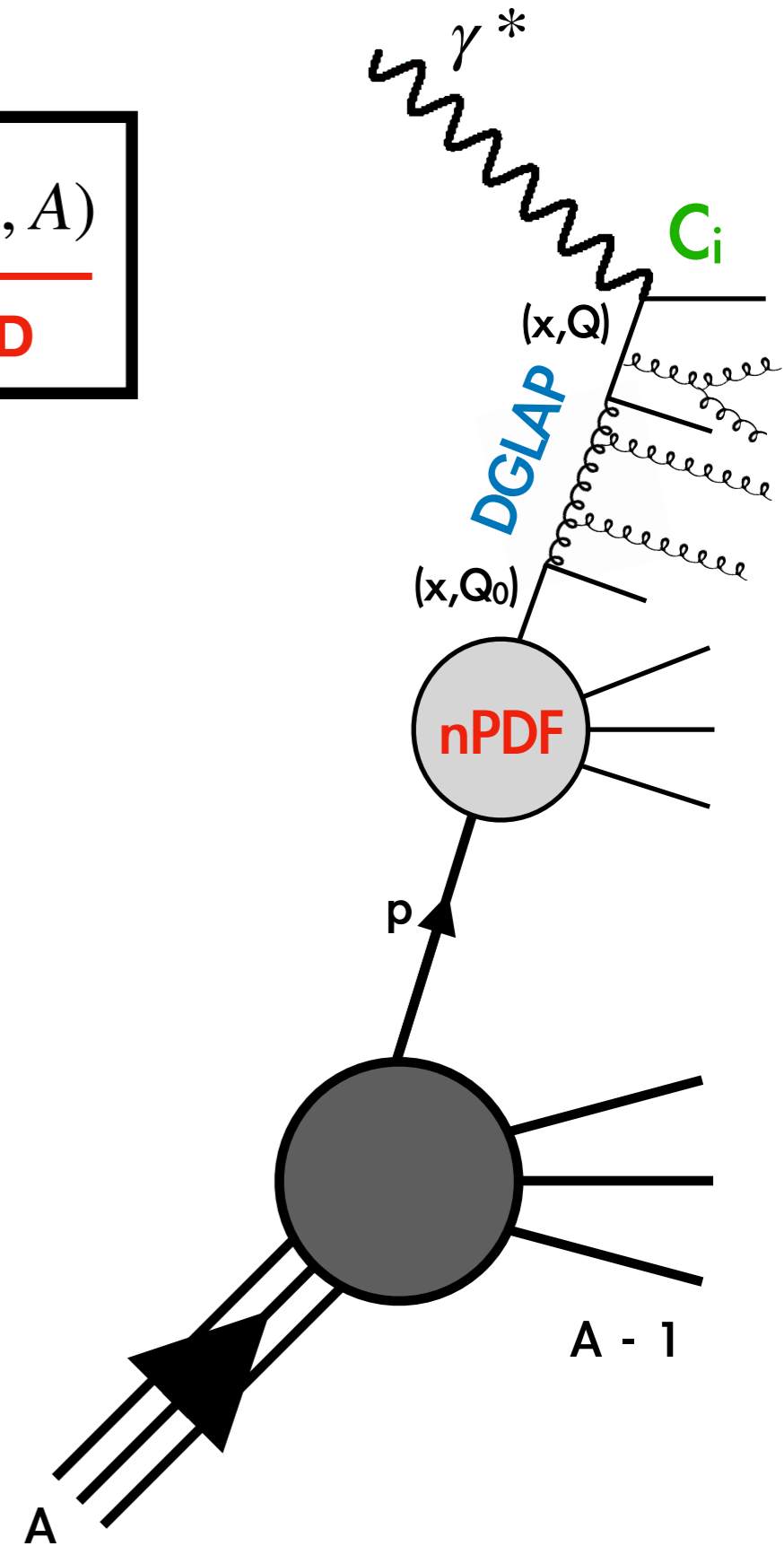
# Data

	EPS09	DSSZ12	KA15	NCTEQ15	EPPS16	nNNPDF1.0
Order in $\alpha_s$	LO & NLO	NLO	NNLO	NLO	NLO	NLO
Neutral current DIS $\ell+A/\ell+d$	✓	✓	✓	✓	✓	✓
Drell-Yan dilepton p+A/p+d	✓	✓	✓	✓	✓	
RHIC pions d+Au/p+p	✓	✓		✓	✓	
Neutrino-nucleus DIS		✓			✓	
Drell-Yan dilepton $\pi+A$					✓	
LHC p+Pb jet data					✓	
LHC p+Pb W, Z data					✓	
$Q$ cut in DIS	1.3 GeV	1 GeV	1 GeV	2 GeV	1.3 GeV	1.3 GeV
datapoints	929	1579	1479	708	1811	605
free parameters	15	25	16	17	20	73
error analysis	Hessian	Hessian	Hessian	Hessian	Hessian	Monte Carlo rep
error tolerance $\Delta\chi^2$	50	30	not given	35	52	NNPDF3.1
Free proton baseline PDFs	CTEQ6.1	MSTW2008	JR09	CTEQ6M-like	CT14NLO	
Heavy-quark effects		✓		✓	✓	✓
Flavor separation				some	✓	
Reference	[JHEP 0904 065]	[PR D85 074028]	[PR D93, 014026]	[PR D93 085037]	[EPJ C77 163]	<b>Preliminary</b>

# Nuclear NC Inclusive DIS

EM  $F_2$  ( $Q < M_Z$ )

$$F_2(x, Q^2, A) = \sum_i^{n_f} \sum_j^{n_f} \underbrace{C_i(x, Q^2)}_{\text{pQCD}} \otimes \underbrace{\Gamma_{ij}(x, Q^2)}_{\text{DGLAP}} \otimes \underbrace{q_j(x, Q_0^2, A)}_{\text{npQCD}}$$



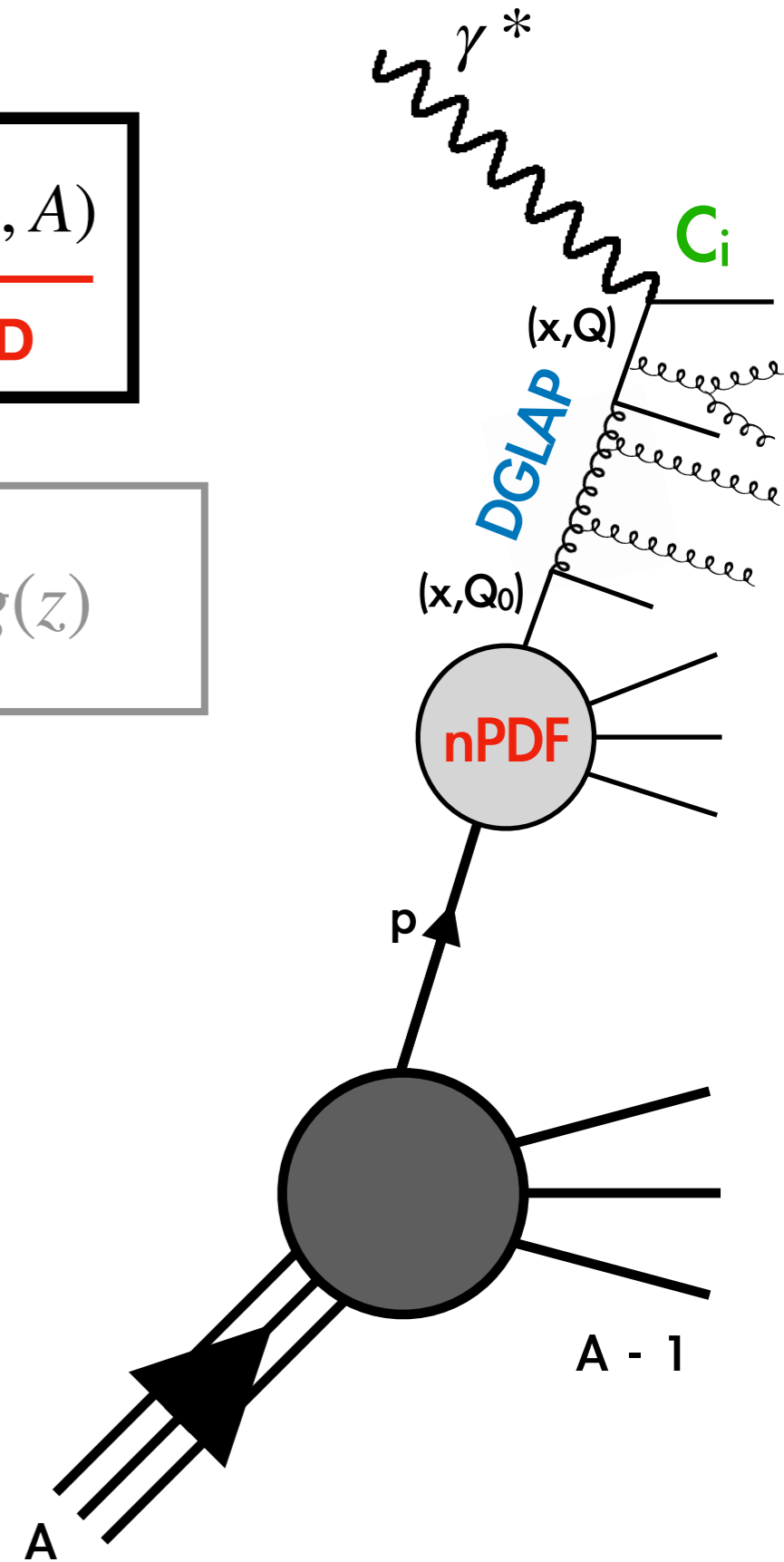
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Convolution

$$f \otimes g = \int_x^1 \frac{dz}{z} f\left(\frac{x}{z}\right) g(z)$$



# Nuclear NC Inclusive DIS

**EM  $F_2$  ( $Q < M_Z$ )**

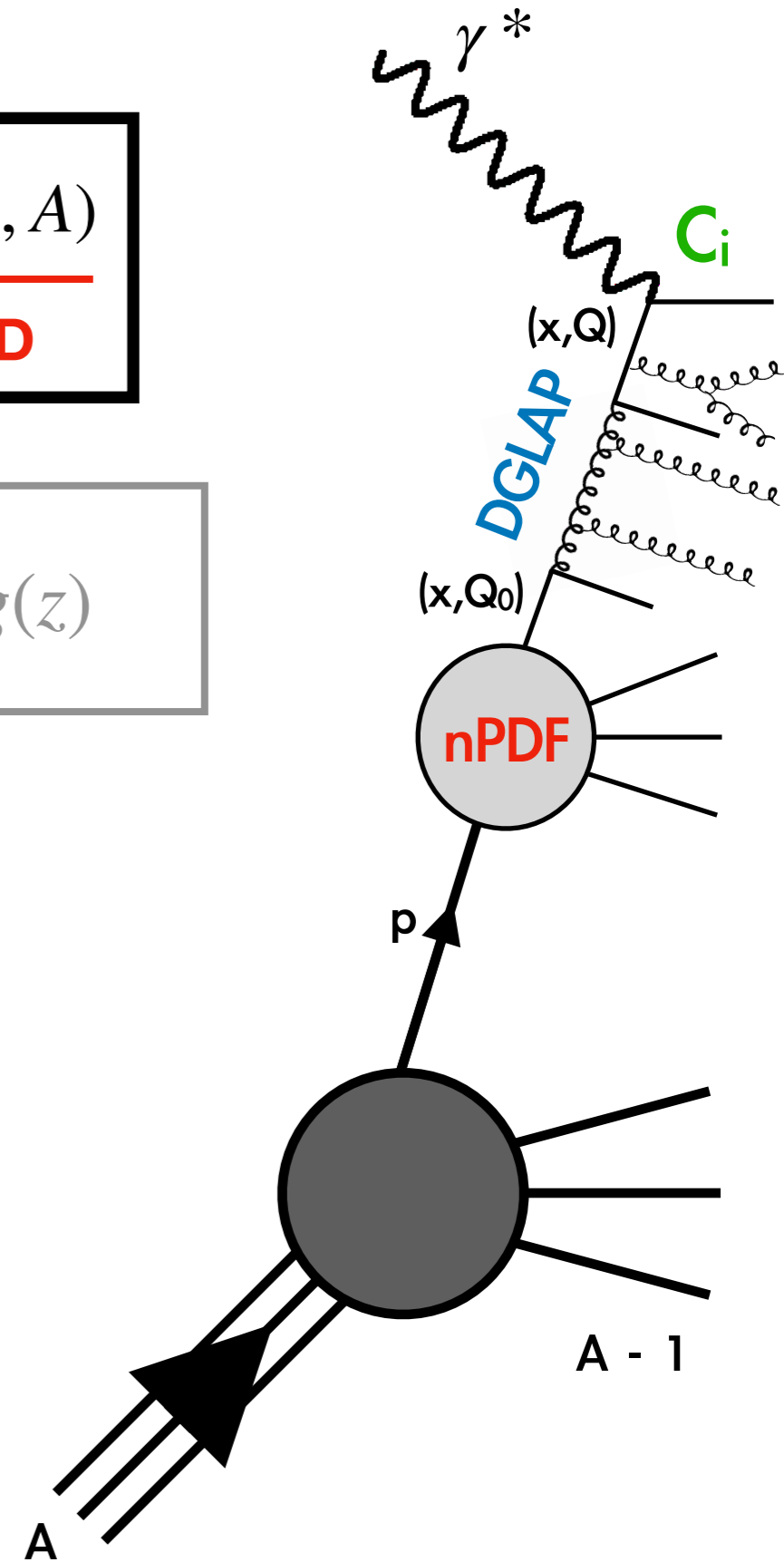
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**Coefficient Functions**

$$C_i = C_i^{(0)} + \frac{\alpha_s}{4\pi} C_i^{(1)} + O(\alpha_s^2)$$

**Convolution**

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# Nuclear NC Inclusive DIS

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## Coefficient Functions

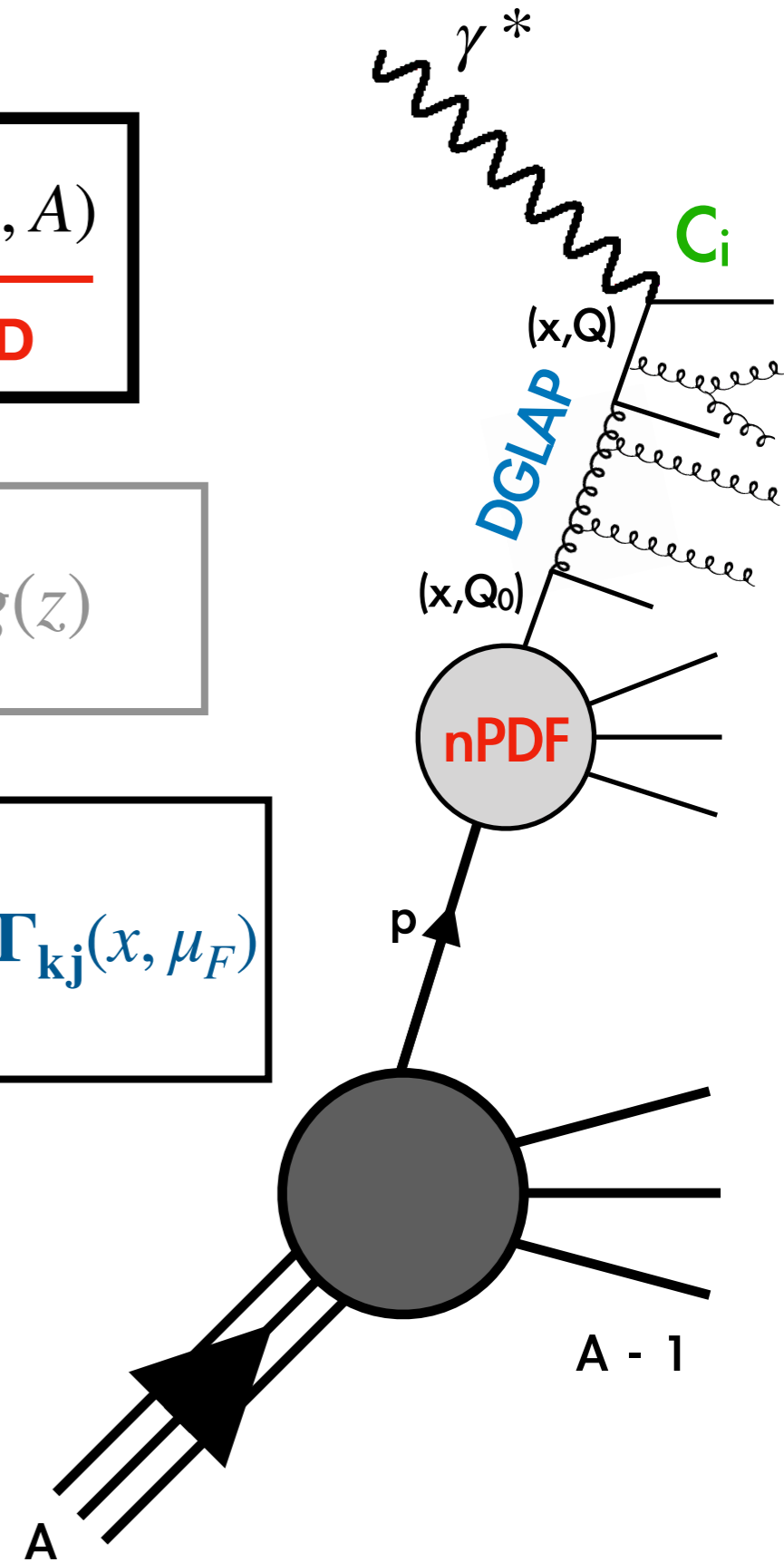
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## Convolution

$$f \otimes g = \int_x^1 \frac{dz}{z} f\left(\frac{x}{z}\right) g(z)$$

## DGLAP equation

$$\frac{\partial \Gamma_{ij}}{\partial \ln \mu_F^2} = \sum_k \frac{\alpha_s(\mu_F)}{4\pi} \left[ P_{ik}^{(0)}(x) + \frac{\alpha_s(\mu_F)}{4\pi} P_{ik}^{(1)}(x) + \dots \right] \otimes \Gamma_{kj}(x, \mu_F)$$





# Nuclear NC Inclusive DIS

## EM $F_2$ ( $Q < M_Z$ )

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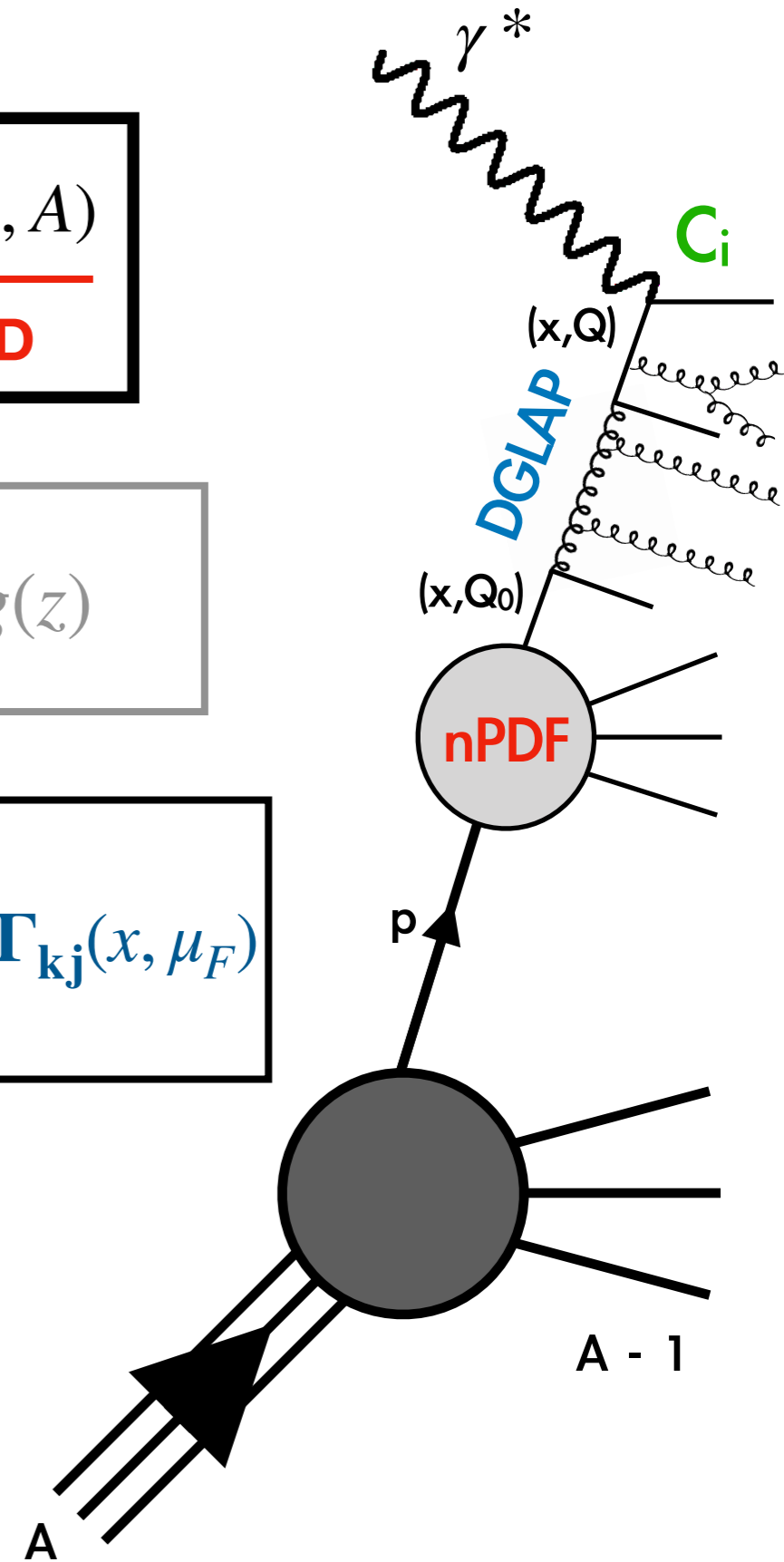
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## RG equation

$$\frac{\partial \alpha_s}{4\pi \cdot \partial \ln \mu_R^2} = - \left( \frac{\alpha_s(\mu_R)}{4\pi} \right)^2 \left[ \beta_0 + \frac{\alpha_s(\mu_R)}{4\pi} \beta_1 + \dots \right]$$



# Nuclear NC Inclusive DIS

**EM  $F_2$  ( $Q < M_Z$ )**

$$F_2(x, Q^2, A) = \sum_i^{n_f} \sum_j^{n_f} \underbrace{C_i(x, Q^2)}_{\text{pQCD}} \otimes \underbrace{\Gamma_{ij}(x, Q^2)}_{\text{DGLAP}} \otimes \underbrace{q_j(x, Q_0^2, A)}_{\text{npQCD}}$$

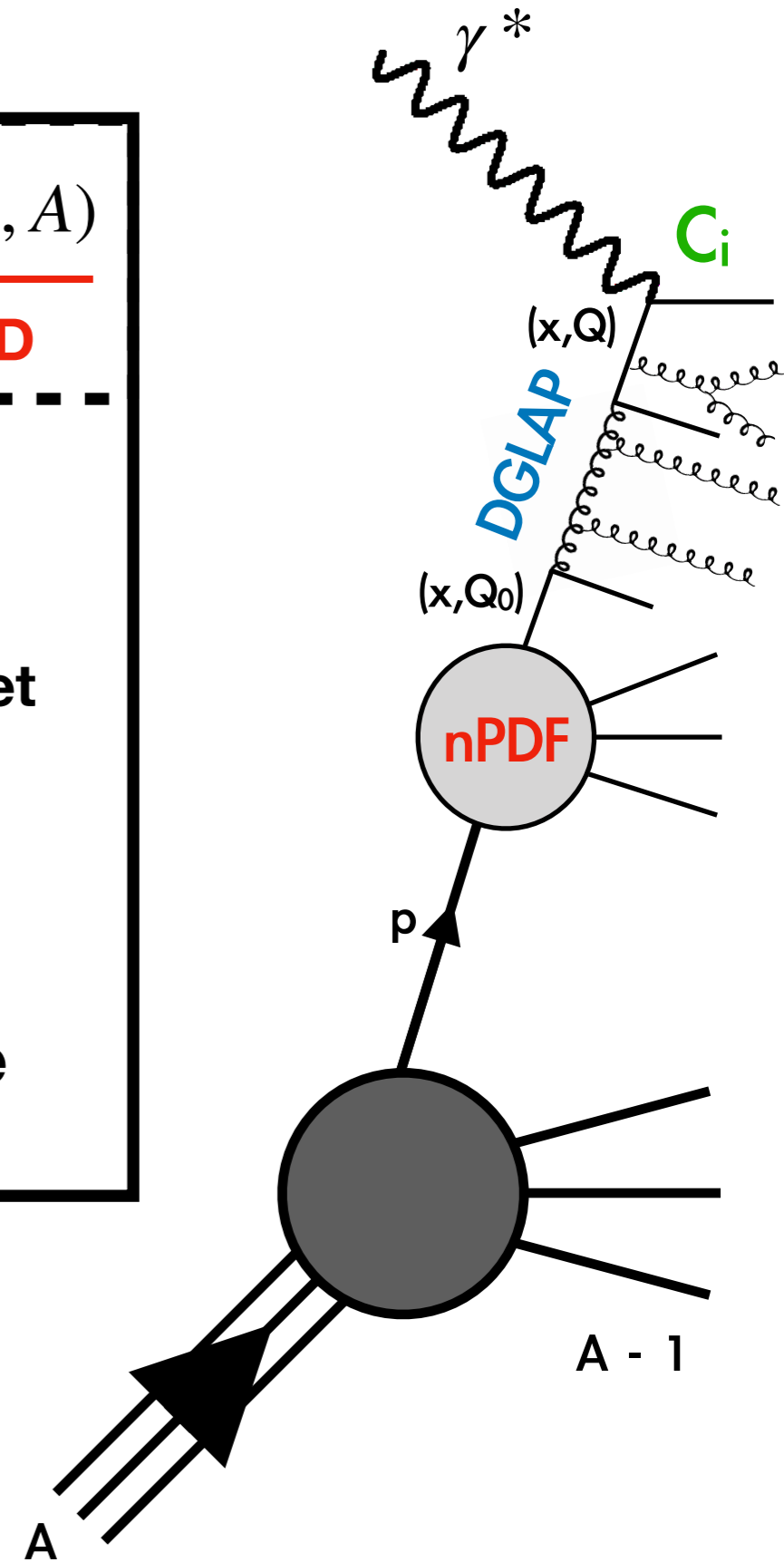
$$= C_{2,q}^S(x, \alpha_s(Q)) \otimes \Sigma(x, Q) \rightarrow \text{Singlet}$$

$$+ C_{2,q}^{NS}(x, \alpha_s(Q)) \otimes T(x, Q^2) \rightarrow \text{Non-Singlet}$$

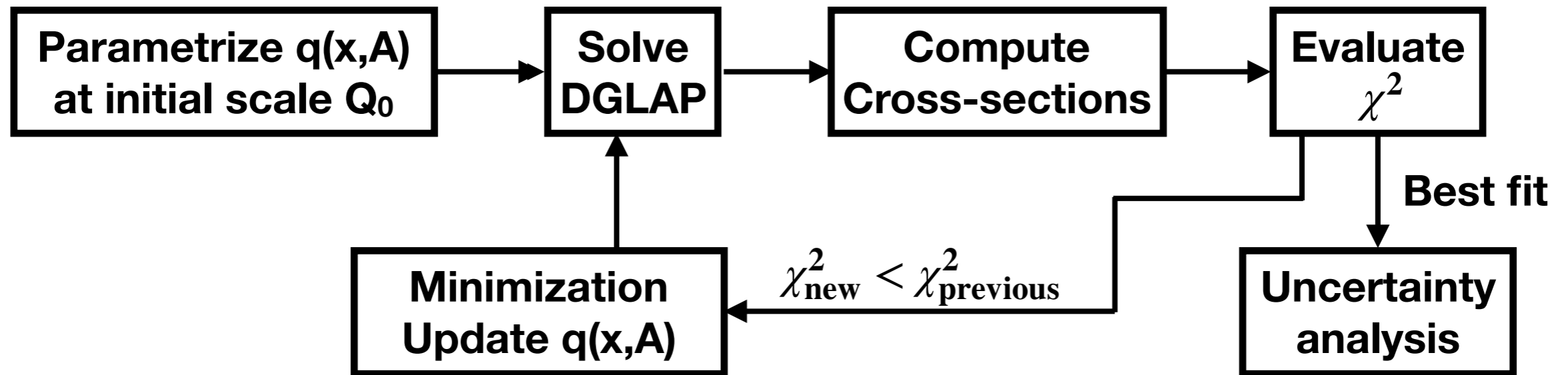
$$+ C_{2,g}^S(x, \alpha_s(Q)) \otimes g(x, Q^2) \rightarrow \text{Gluon}$$

Theoretically, 3 Independent PDFs [backup]

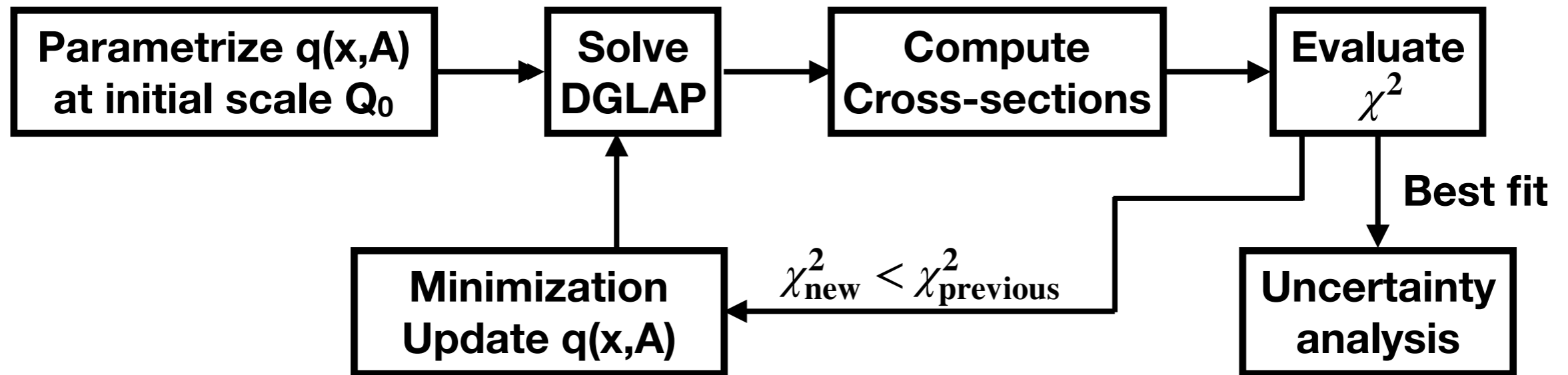
Practically, depends on data kinematic range



# Methodology

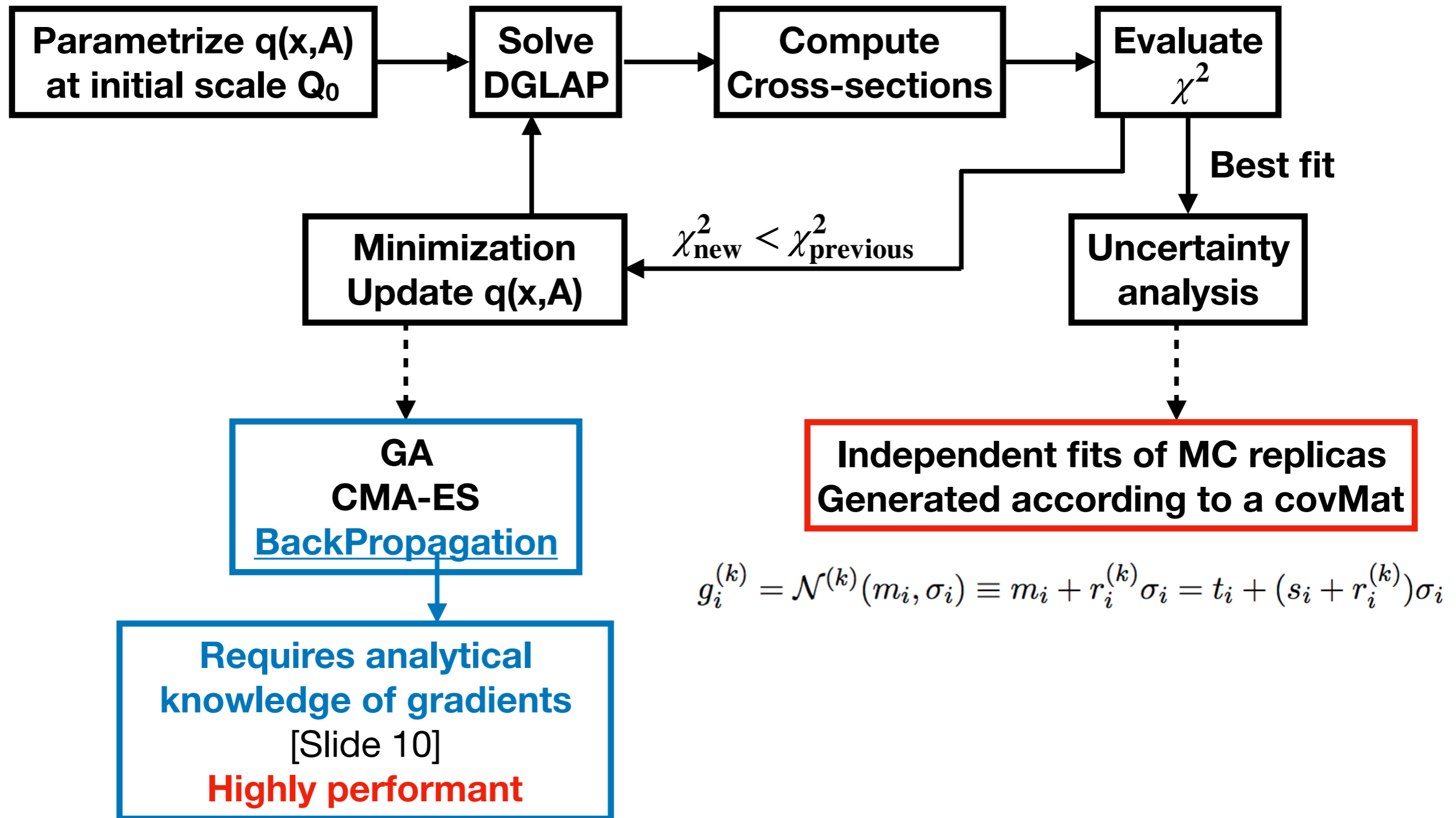


# Methodology



$$\chi^2 = \frac{1}{N_{data}} \sum_{n,m} (\hat{F}_2^{(n)} - F_2^{(n)}) Cov_{nm}^{-1} (\hat{F}_2^{(m)} - F_2^{(m)})$$

# Methodology



# Status of recent nPDFs

## EPPS16 [1612.05741]

$$f_i^{P/A}(x, Q^2) = \underline{R_i^A}(x, Q^2) f_i^P(x, Q^2)$$

$$R_i^A(x, Q_0^2) = \begin{cases} a_0 + a_1(x - x_a)^2 & x \leq x_a \\ b_0 + b_1 x^\alpha + b_2 x^{2\alpha} + b_3 x^{3\alpha} & x_a \leq x \leq x_e \\ c_0 + (c_1 - c_2 x)(1 - x)^{-\beta} & x_e \leq x \leq 1, \end{cases}$$

## nCTEQ15 [1509.00792]

$$x f_i^{p/A}(x, Q_0) = c_0 x^{c_1} (1 - x)^{c_2} e^{c_3 x} (1 + e^{c_4 x})^{c_5},$$

for  $i = u_v, d_v, g, \bar{u} + \bar{d}, s + \bar{s}, s - \bar{s}$ ,

$$\frac{\bar{d}(x, Q_0)}{\bar{u}(x, Q_0)} = c_0 x^{c_1} (1 - x)^{c_2} + (1 + c_3 x)(1 - x)^{c_4}.$$

$$c_k \rightarrow c_k(A) \equiv c_{k,0} + c_{k,1} (1 - A^{-c_{k,2}}),$$

$k = \{1, \dots, 5\}.$

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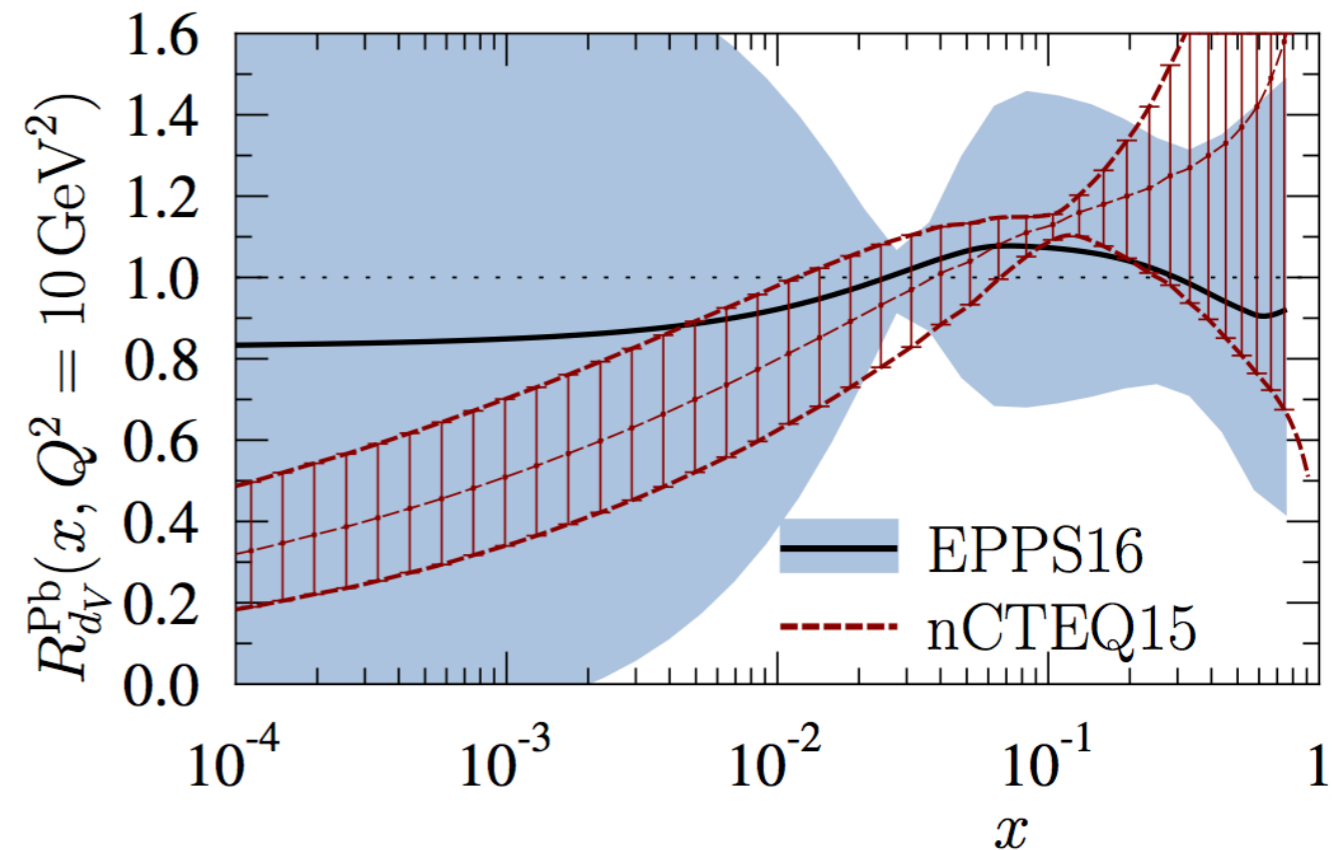
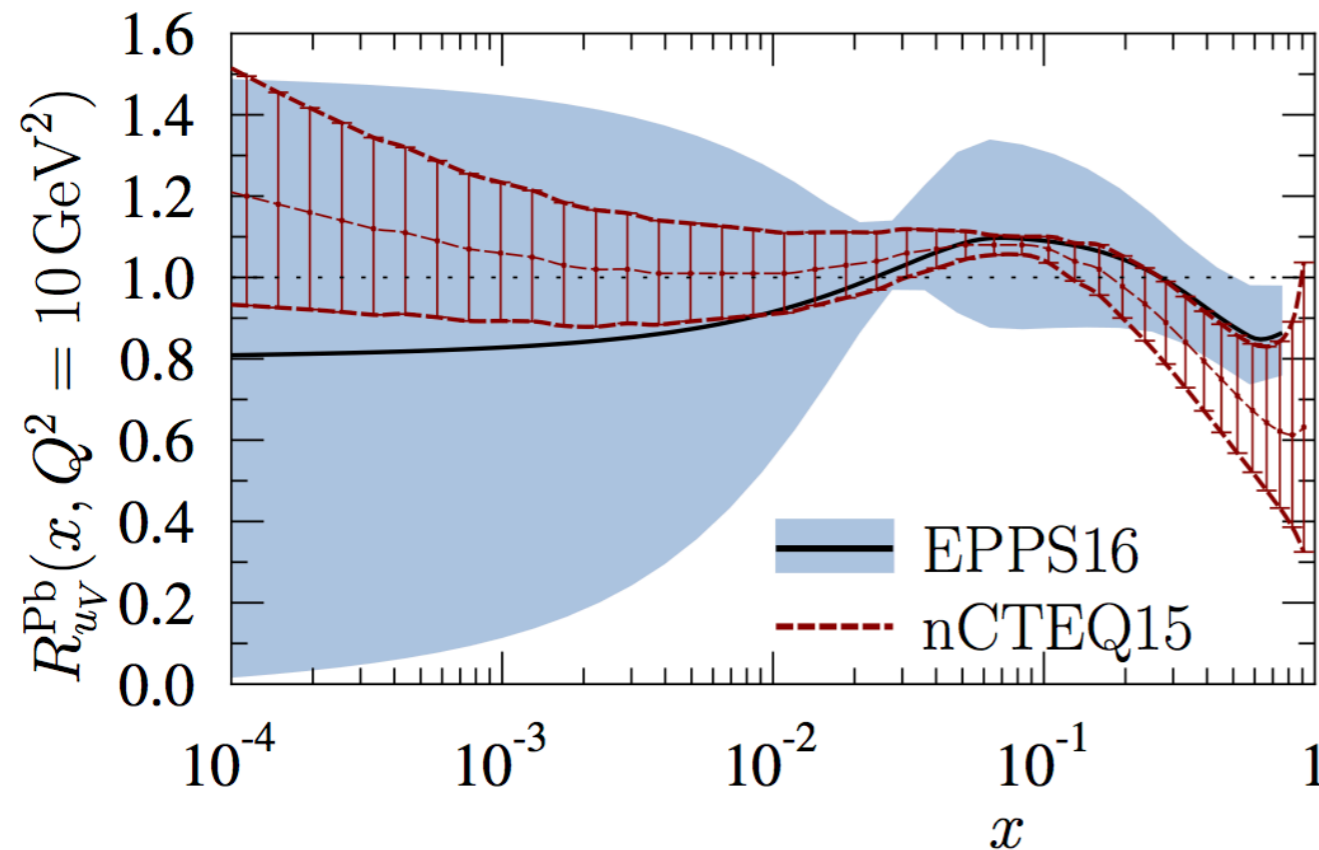
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# Parametrizations

**EM  $F_2$  ( $Q < M_Z$ )**

$$\begin{aligned} F_2(x, Q^2, A) &= \sum_i^{n_f} \sum_j^{n_f} C_i(x, Q^2) \otimes \Gamma_{ij}(x, Q^2) \otimes q_j(x, Q_0^2, A) \\ &= \sum_i^{n_f} \widetilde{C}_i(x, Q^2) \otimes \mathbf{q}_i(\mathbf{x}, Q_0^2, A) \end{aligned}$$



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**Pre-computed FK tables  
[NNPDF methodology]**

# Parametrizations

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## PDF - Proton case

$$q_i^p(x, Q_0^2) = x f_i(x, Q_0^2) \propto \mathbf{x}^\alpha (1 - \mathbf{x})^\beta \mathbf{NN}(\mathbf{x})$$

# Parametrizations

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## Option A

$$\mathbf{G}(\mathbf{x}, \mathbf{A}, \mathbf{Q}_0^2) = A_G \cdot (1 - x)^{\alpha_G} \mathbf{NN}_G(\mathbf{x}, \mathbf{A})$$

$$\mathbf{\Sigma}(\mathbf{x}, \mathbf{A}, \mathbf{Q}_0^2) = (1 - x)^{\alpha_\Sigma} \cdot \mathbf{NN}_\Sigma(\mathbf{x}, \mathbf{A})$$

$$\mathbf{T}_8 = (1 - x)^{\alpha_{T8}} \cdot \mathbf{NN}_{T8}(\mathbf{x}, \mathbf{A})$$

$$\text{Where } A_G = \frac{1 - \int_0^1 x \Sigma(x, \mathbf{A}) dx}{\int_0^1 x G(x, \mathbf{A}) dx}$$

# Parametrizations

## Option B

$$q_i(x, A, Q_0^2) = \frac{1}{SR} [\text{NN}_i(x, A, Q_0^2) - \text{NN}_i(x=1, A, Q_0^2)]$$

$$\text{Where } SR = \int_0^1 x [\text{NN}_G(x, A) + \text{NN}_\Sigma(x, A)] dx$$

## Option A

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$$T_8 = (1-x)^{\alpha_{T8}} \cdot \text{NN}_{T8}(x, A)$$

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# Parametrizations

## Option B

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$$\text{Where } SR = \int_0^1 x [\text{NN}_G(x, A) + \text{NN}_\Sigma(x, A)] dx$$

**Next Results are for:**

## Option A

$$G(\mathbf{x}, A, Q_0^2) = A_G \cdot (1 - x)^{\alpha_G} \text{NN}_G(\mathbf{x}, A)$$

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# Neural Network

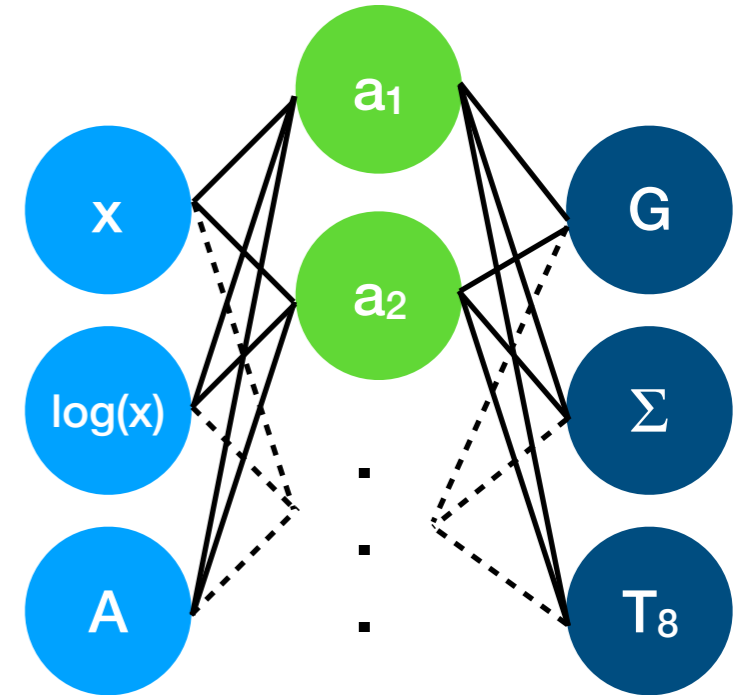
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$$\mathbf{T}_8 = (1 - x)^{\alpha_{T8}} \cdot \mathbf{NN}_{T8}(\mathbf{x}, \mathbf{A})$$

$$\text{Where } A_G = \frac{1 - \int_0^1 x \Sigma(x, \mathbf{A}) dx}{\int_0^1 x G(x, \mathbf{A}) dx}$$



## Neural Network

$$\mathbf{NN}_i(\mathbf{x}; \{\mathbf{w}_{ij}^{(\ell)}, \mathbf{b}_i^{(\ell)}\}) = \sigma_L \left( \sum_j^{N_{L-1}} w_{ij}^{(L)} a_j^{(L-1)} + b_i^{(L)} \right)$$

$$= \sigma_L \left( \sum_j^{N_{L-1}} w_{ij}^{(L)} \sigma_{L-1} \left( \sum_k^{N_{L-2}} w_{jk}^{(L-1)} a_k^{(L-2)} + b_j^{(L-1)} \right) + b_i^{(L)} \right) = \dots$$

# Backpropagation

## Gradient Descent

$$w_{ij} = w_{ij} - \frac{\eta}{N_{data}} \sum_n \frac{\partial \chi^{2(n)}}{\partial w_{ij}}$$

$$b_i = b_i - \frac{\eta}{N_{data}} \sum_n \frac{\partial \chi^{2(n)}}{\partial b_i}$$

## $\chi^2$ Derivatives

$$\frac{\partial \chi^2}{\partial w_{mn}^{(\ell)}} = 2 \left( \frac{\tilde{\mathbf{C}} \otimes \mathbf{NN} - F}{\sigma^2} \right) \tilde{\mathbf{C}} \otimes \frac{\partial \mathbf{NN}}{\partial w_{mn}^{(\ell)}}$$

$$\frac{\partial \chi^2}{\partial b_m^{(\ell)}} = 2 \left( \frac{\tilde{\mathbf{C}} \otimes \mathbf{NN} - F}{\sigma^2} \right) \tilde{\mathbf{C}} \otimes \frac{\partial \mathbf{NN}}{\partial b_m^{(\ell)}}$$

## NN Derivatives

$$\frac{\partial \mathbf{NN}}{\partial w_{ij}^{(\ell)}} = \left[ \prod_{\alpha=L}^{\ell+1} \mathbf{W}^{(\alpha)} \right] \cdot \frac{\partial \mathbf{a}^{(\ell)}}{\partial w_{ij}^{(\ell)}}$$

$$\frac{\partial \mathbf{NN}}{\partial b_i^{(\ell)}} = \left[ \prod_{\alpha=L}^{\ell+1} \mathbf{W}^{(\alpha)} \right] \cdot \frac{\partial \mathbf{a}^{(\ell)}}{\partial b_i^{(\ell)}}$$

## Neurons Derivatives

$$\frac{\partial \mathbf{a}_k^{(\ell)}}{\partial w_{ij}^{(\ell)}} = \sigma'_\ell \left( z_k^{(\ell)} \right) \delta_{ik} a_j^{(\ell-1)}$$

$$\frac{\partial \mathbf{a}_k^{(\ell)}}{\partial b_i^{(\ell)}} = \sigma'_\ell \left( z_k^{(\ell)} \right) \delta_{ik}$$

# Results: Theory Benchmark

## FK tables computation

EM  $F_2$  ( $Q < M_Z$ )

$$F_2(x, Q^2, A) = \sum_i^{n_f} \widetilde{C}_i(x, Q^2) \otimes \mathbf{q}_i(\mathbf{x}, Q_0^2, A)$$

	DataSet	APFELcomb	EPPS16	Ndata
$\chi^2_{\text{total}}$	NMC (He/D)	17.76	18.0	16/18
	NMC (Li/D)	19.64	18.4	15/24
	NMC (C/D)	25.55	25.70	31/42
	NMC (Ca/D)	28.49	27.6	15/18



# Results: WarmUp

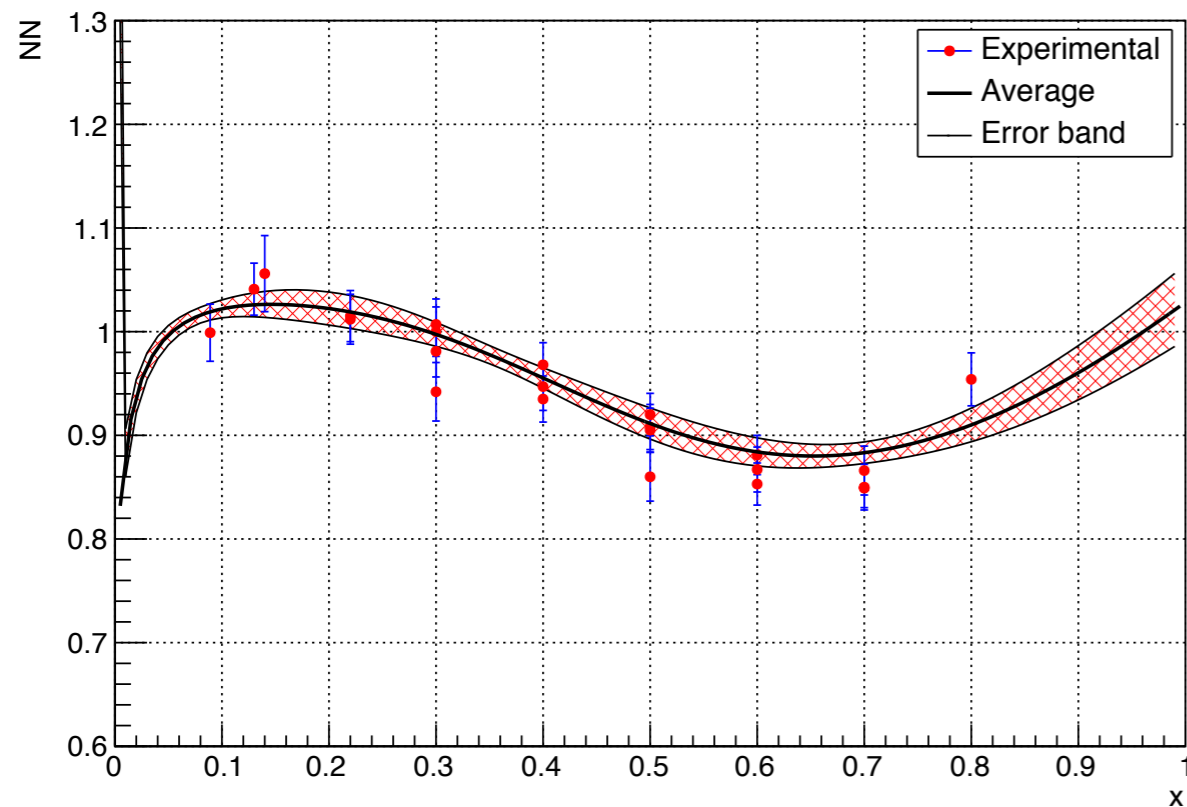
## $F_2^A/F_2^D$ Ratio Fit

### Analysis

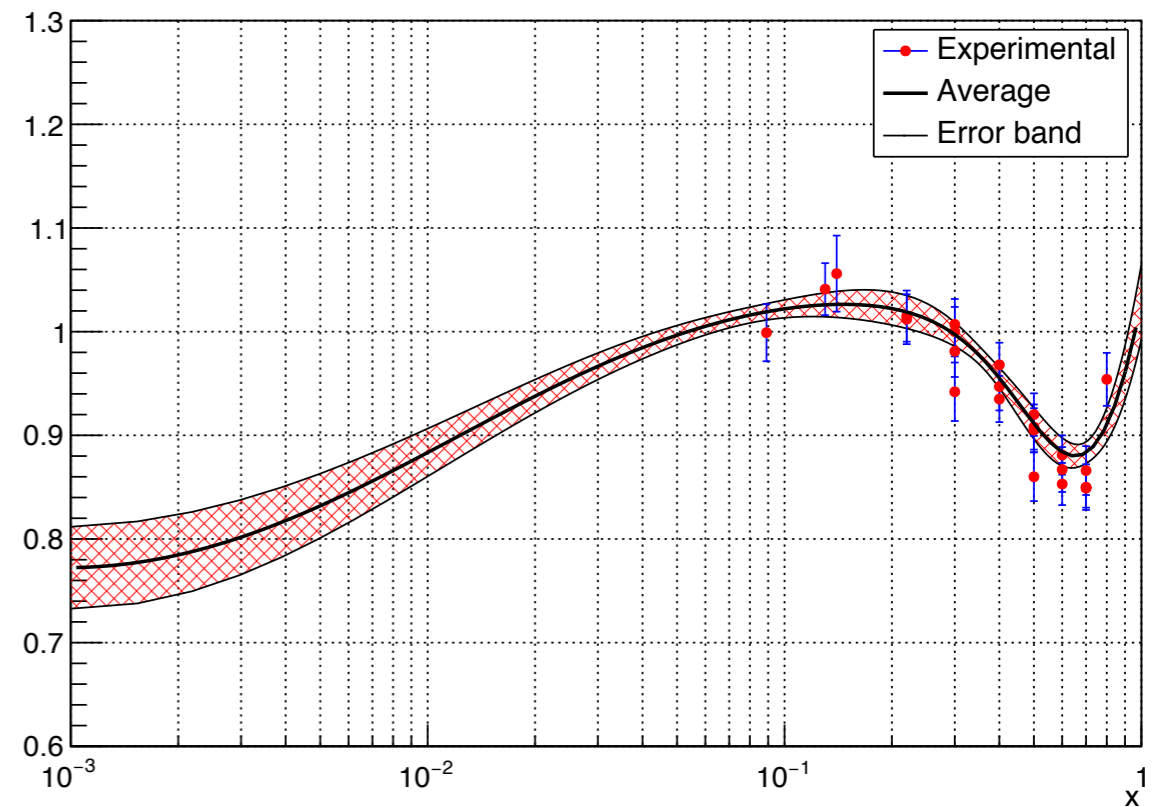
$$NN(x, A) = \frac{F_2^A(x)}{F_2^D(x)}$$

2D fit in (A, x) - Uncorrelated uncertainty

For A = 56

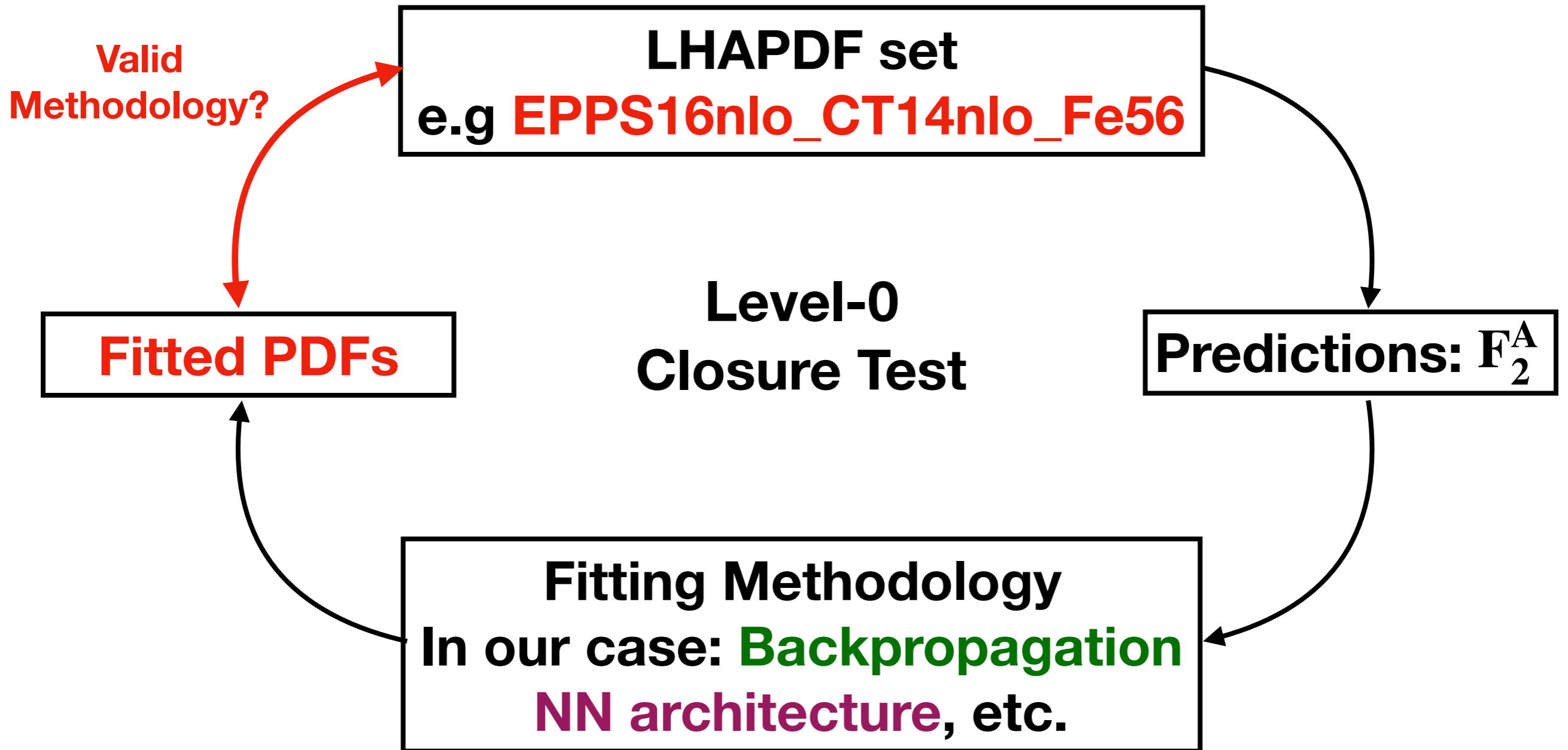


For A = 56



# Results: Closure Test NNPDF3.0 [1410.8849]

## Concept



# Results: Closure Test

NNPDF3.0  
[1410.8849]

## nPDF Fits: Observable Comparison

### Analysis

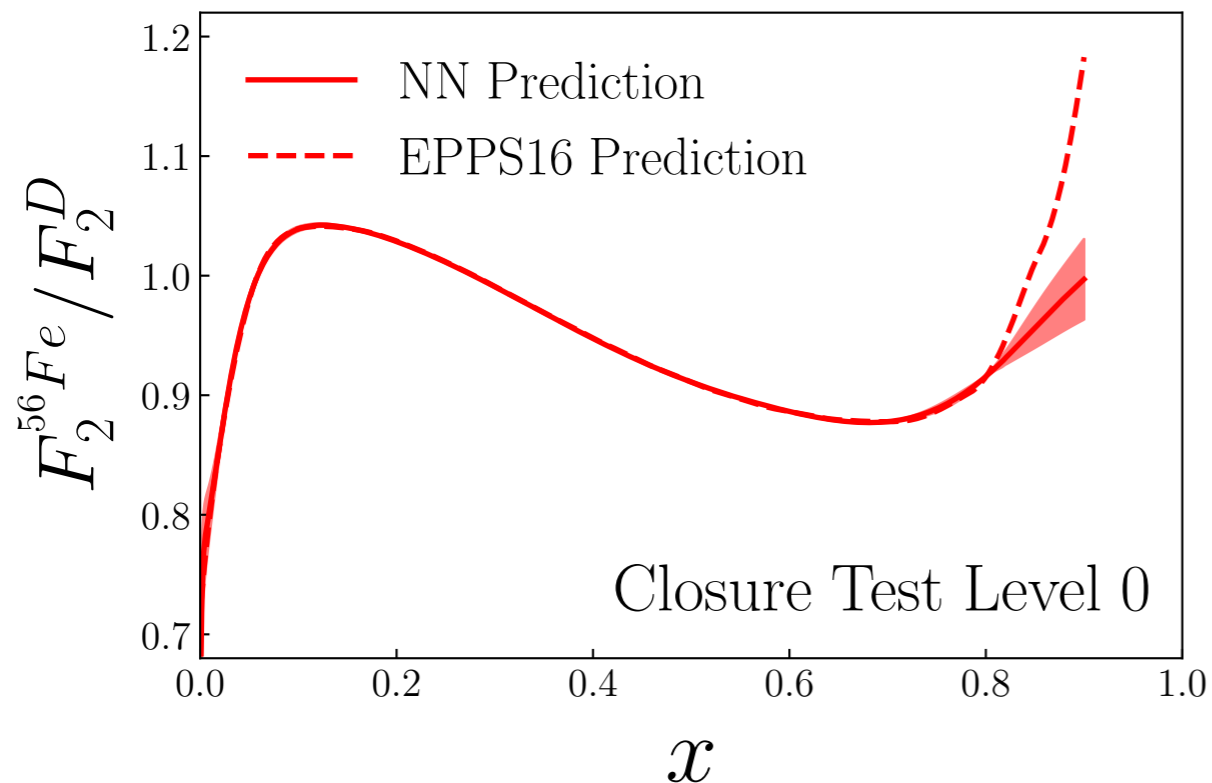
Predictions  $\frac{F_2^A}{F_2^D} \rightarrow$  EPPS16  
 $\frac{F_2^A}{F_2^D} \rightarrow$  CT14nlo

Same theory  
Mass scheme, quarks mass...

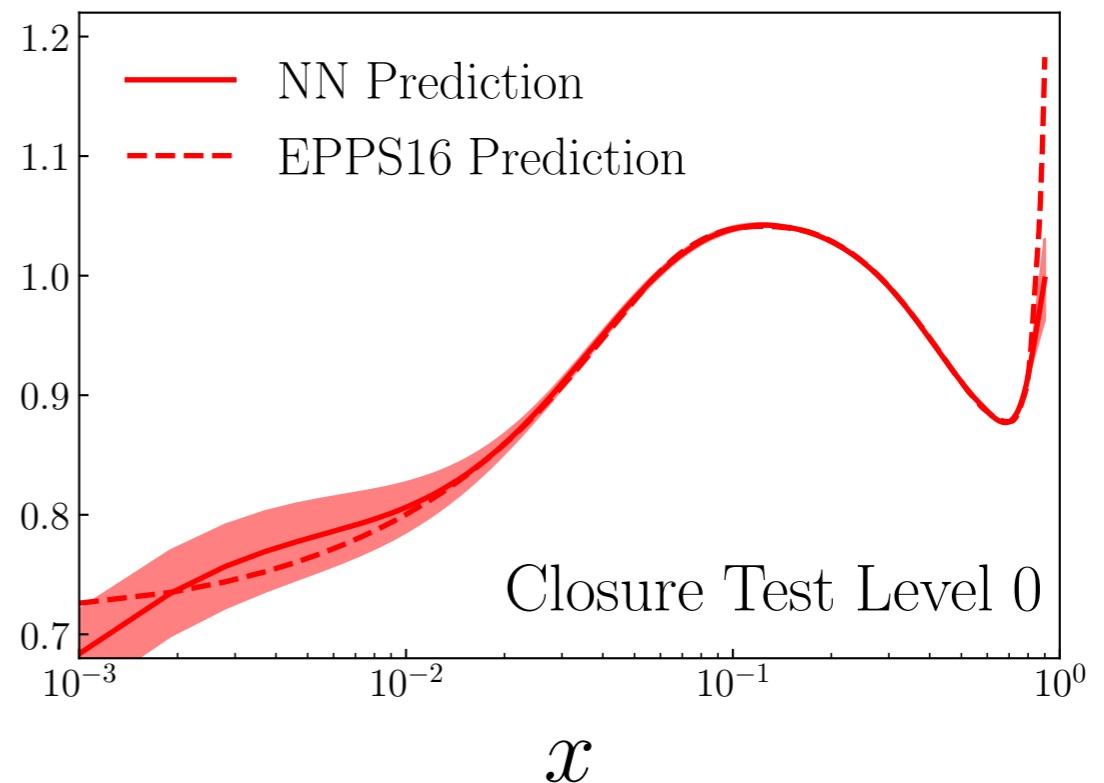
5 different Fits (TensorFlow + ADAM optimiser)

### Iron

PRELIMINARY



PRELIMINARY



# Results: Closure Test

NNPDF3.0  
[1410.8849]

## nPDF Fits: Observable Comparison

### Analysis

Predictions  $\frac{F_2^A}{F_2^D} \rightarrow$  EPPS16  
 $\frac{F_2^A}{F_2^D} \rightarrow$  CT14nlo

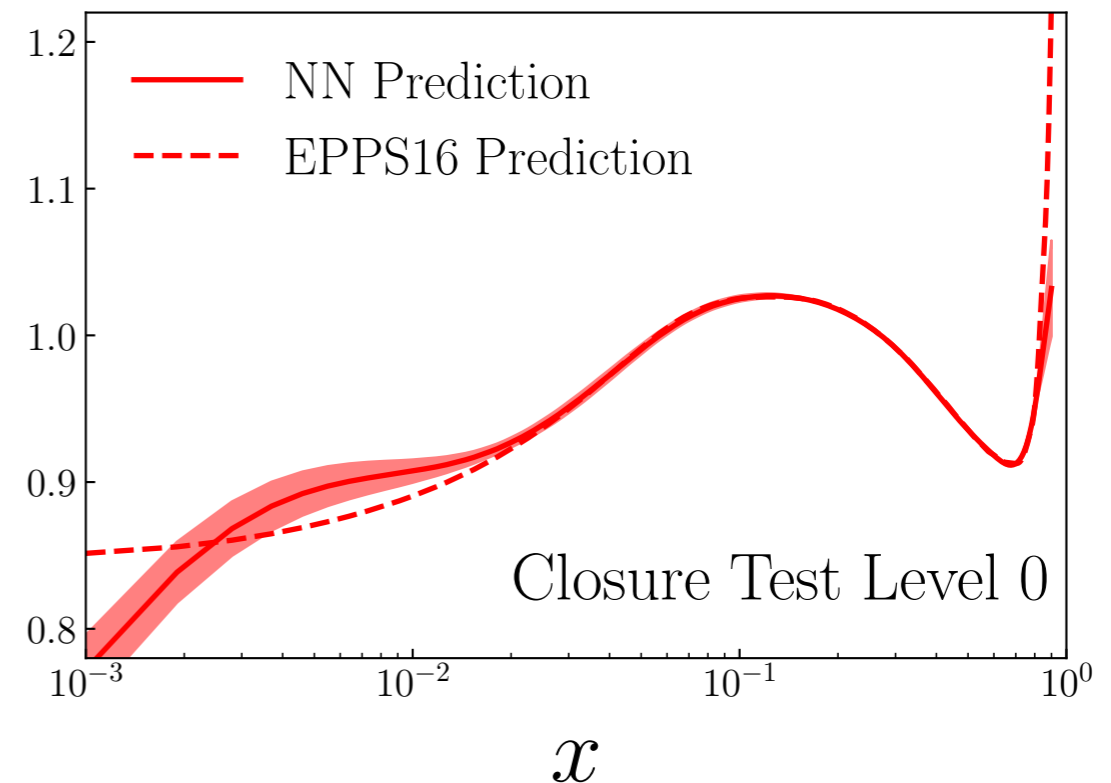
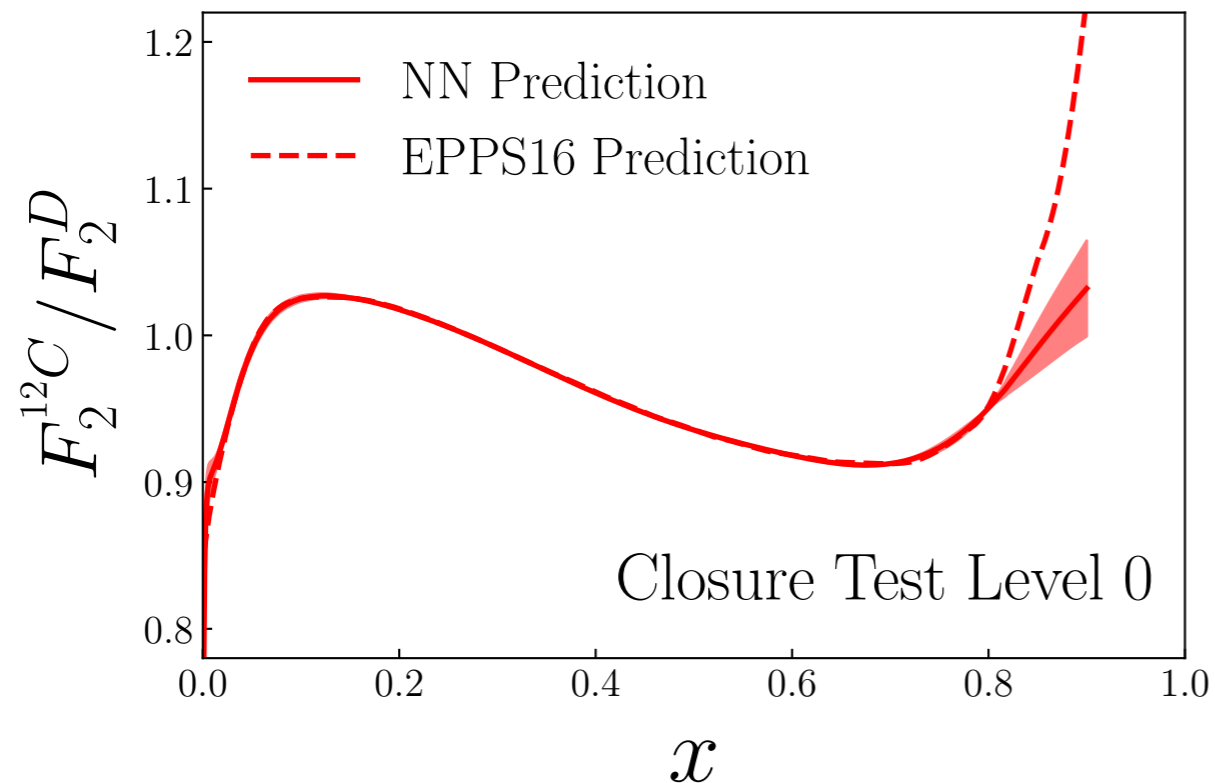
Same theory  
Mass scheme, quarks mass...

5 different Fits (TensorFlow + ADAM optimiser)

### Carbon

PRELIMINARY

PRELIMINARY



# Results: Closure Test

NNPDF3.0  
[1410.8849]

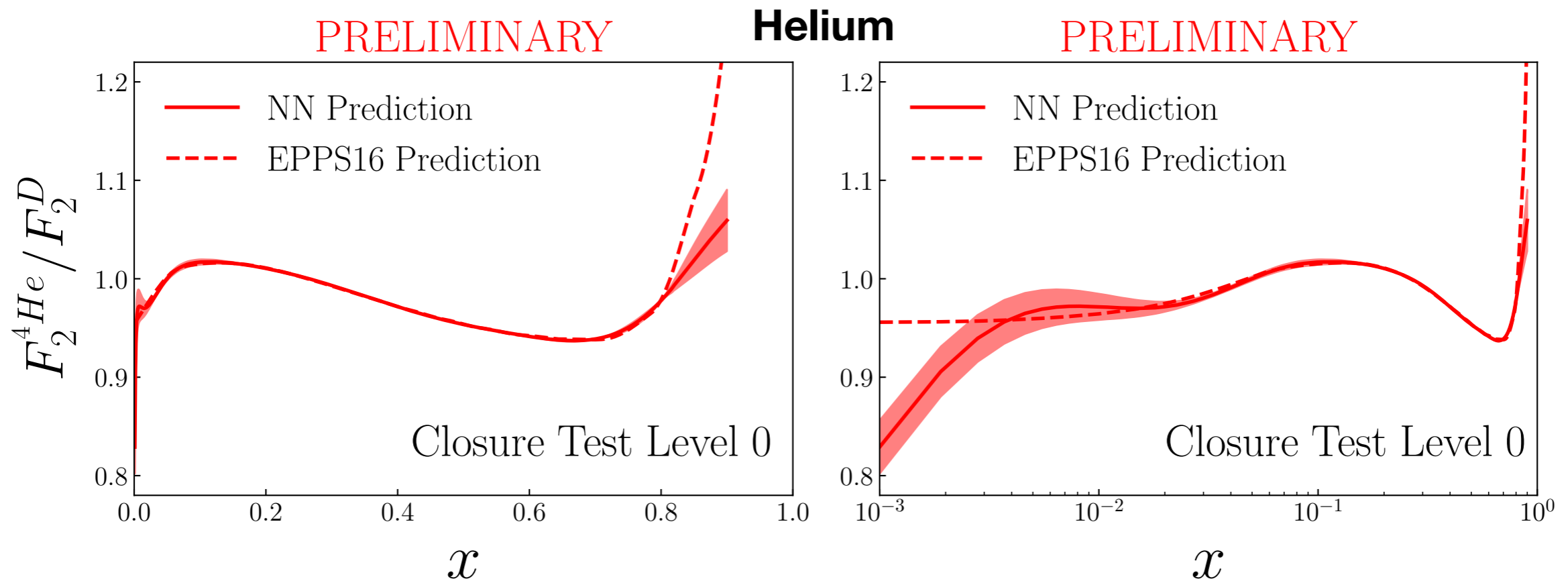
## nPDF Fits: Observable Comparison

### Analysis

Predictions  $\frac{F_2^A}{F_2^D} \rightarrow$  EPPS16  
 $\frac{F_2^A}{F_2^D} \rightarrow$  CT14nlo

Same theory  
Mass scheme, quarks mass...

5 different Fits (TensorFlow + ADAM optimiser)



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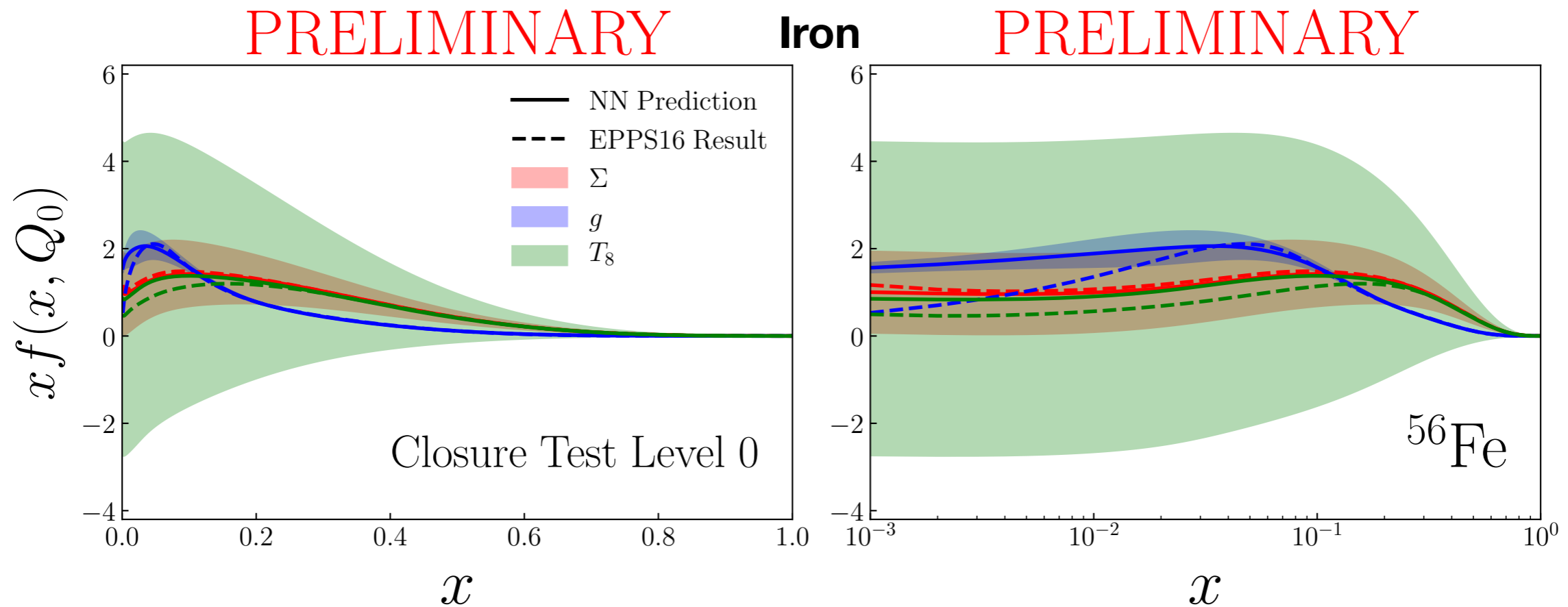
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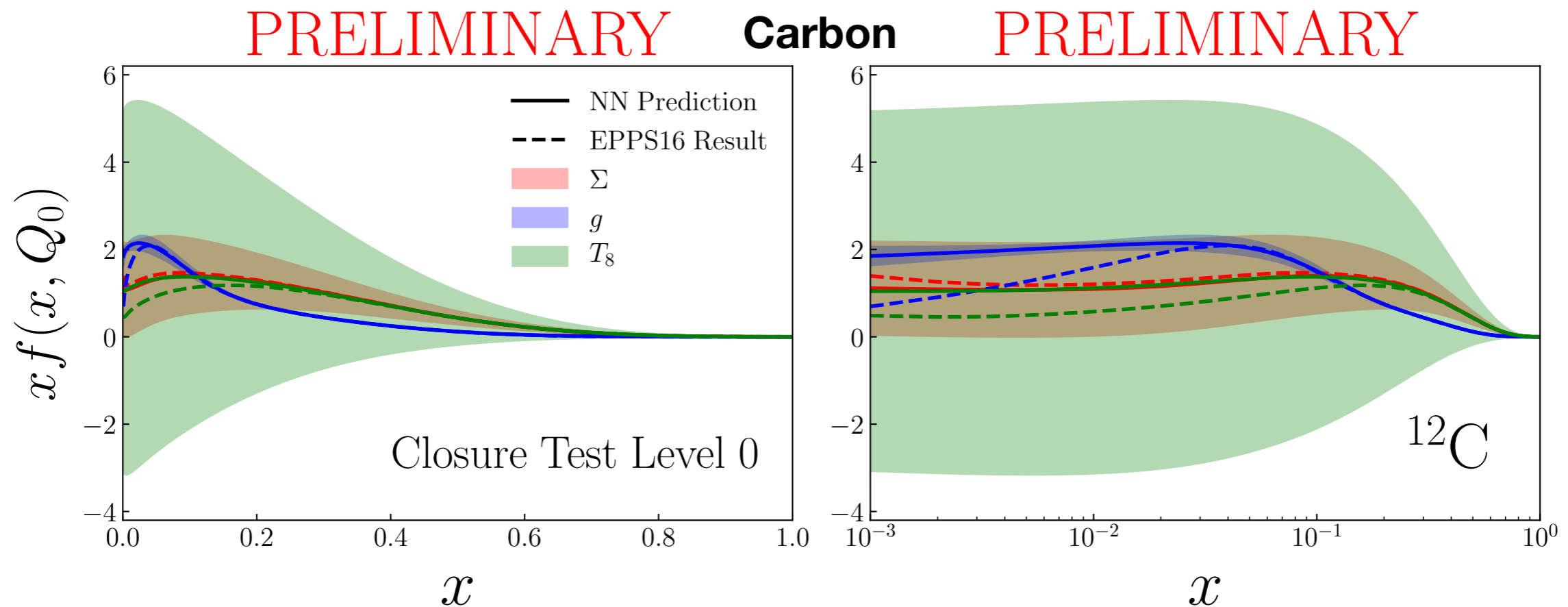
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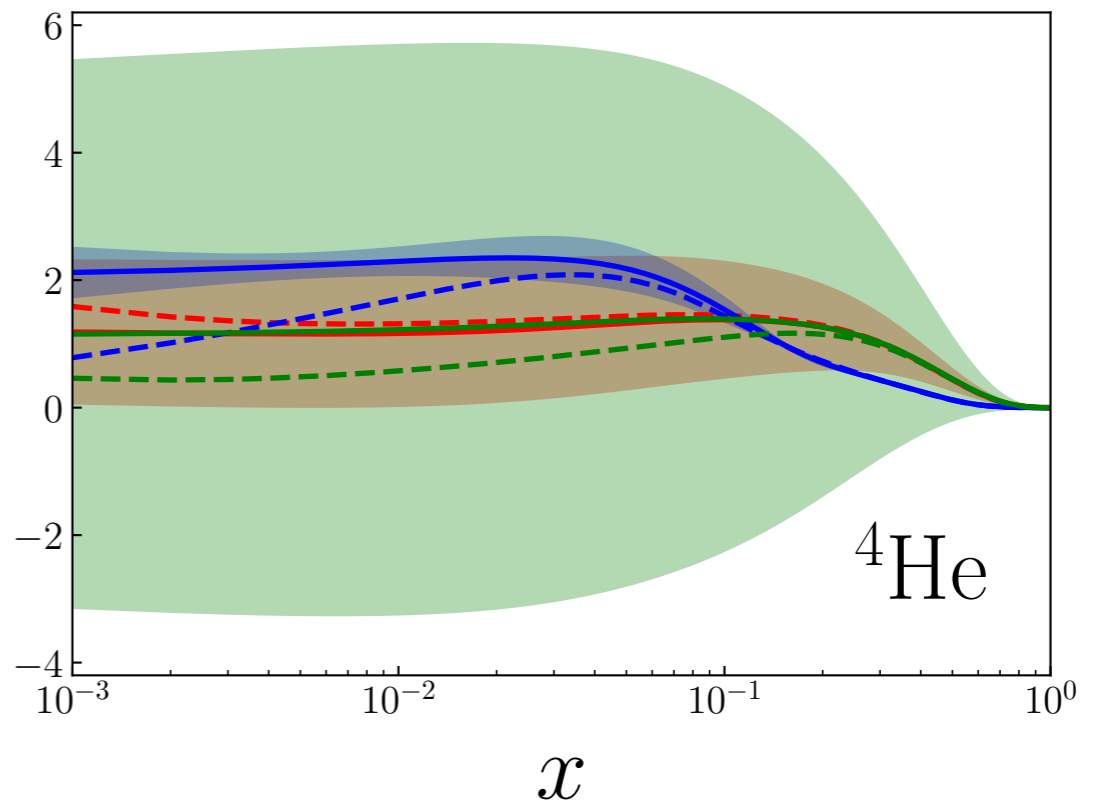
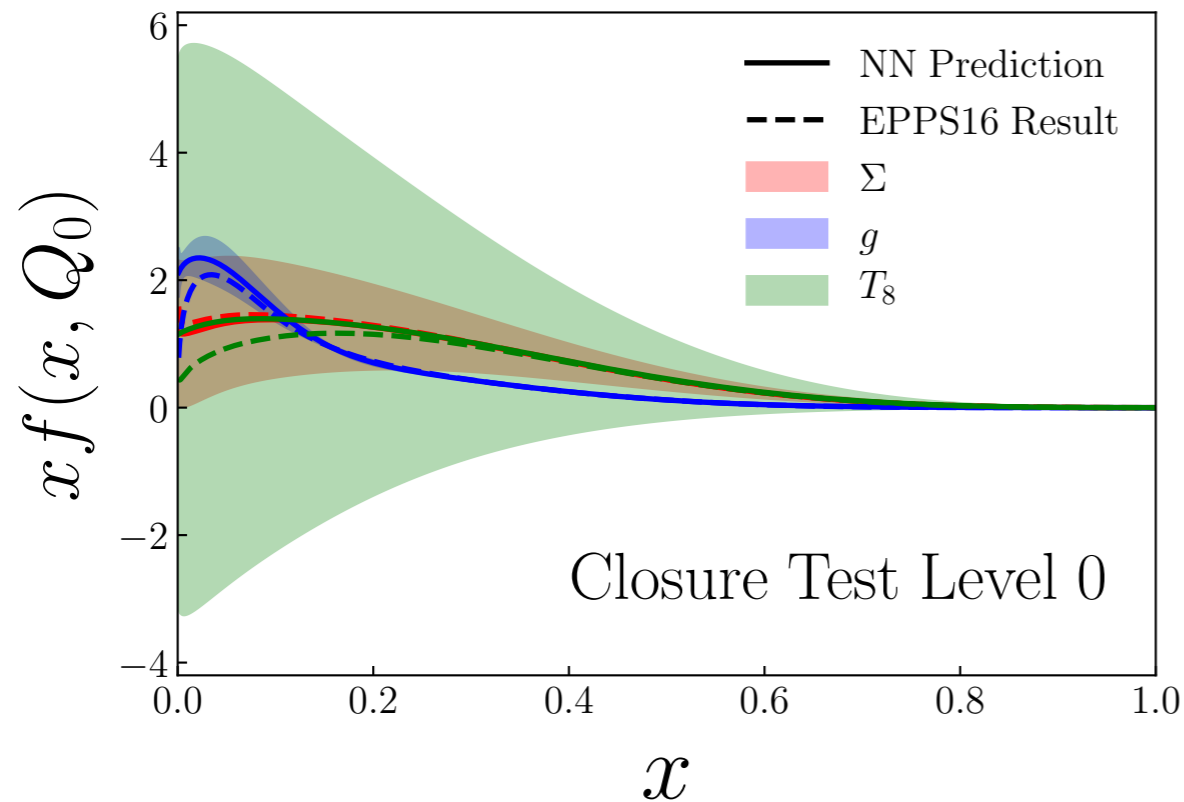
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PRELIMINARY

**Helium**

PRELIMINARY





# Conclusion

- **Theory Benchmarked with EPPS16, On going for nCTEQ15.**
- **Constraints implemented on the level of the fit.**
- **Successful Level 0 closure tests using EPPS16**
  1. First use of Backpropagation to fit PDFs.
  2. Framework in place to perform DIS-only fits.

# Outlook

- **Full-fledged Error analysis:**
  1. Correlated uncertainties.
  2. MC replica generation.
- **Better way of implementing constraints?**
- **Adding DY and additional processes to the fit.**

# Backup

$$F_2(x, Q^2, A) = \sum_i^{n_f} C_i(x, Q^2) \otimes q_i(x, Q^2, A) \quad (2)$$

where  $\otimes$  denotes the Mellin convolution defined as:

$$f(x) \otimes g(x) = \int_0^1 dy \int_0^1 dz f(y)g(z)\delta(x - yz) = \int_x^1 \frac{dy}{y} f(y)g\left(\frac{x}{y}\right) = \int_x^1 \frac{dz}{z} f\left(\frac{x}{z}\right)g(z) \quad (3)$$

We can factorize the nPDF dependence out of the convolution via expansion over a set of interpolating functions, spanning  $Q^2$  and  $x$  such as:

$$q_i(x, Q^2, A) = \sum_{\beta} \sum_{\tau} q_{i,\beta\tau} I_{\beta}(x) I_{\tau}(Q^2) \quad (4)$$

where the nPDF  $q_{i,\beta\tau}$  may be expressed as a product of nPDF at some initial fitting scale  $Q_0$  and an evolution operator obtained by the solution of the DGLAP equation via the interpolation procedure as:

$$q_{i,\beta\tau} \equiv q_i(x_{\beta}, Q_{\tau}^2, A) = \sum_j \sum_{\alpha} \Gamma_{ij,\alpha\beta}^{\tau} q_j(x_{\alpha}, Q_0^2, A) \quad (5)$$

Finally:

$$F_2(x, Q^2, A) = \sum_i^{n_f} C_i(x, Q^2) \otimes \left[ \sum_{\alpha} \sum_{\tau} \sum_j \sum_{\beta} \Gamma_{ij,\alpha\beta}^{\tau} q_j(x_{\alpha}, Q_0^2) I_{\beta}(x) I_{\tau}(Q^2) \right] \quad (6)$$

contracting the sum over  $\tau$ ,  $i$  and  $\beta$  then replacing  $j$  with  $i$  :

$$F_2(x, Q^2) = \sum_i^{n_f} \sum_{\alpha}^{n_x} FK_{i,\alpha}(x, x_{\alpha}, Q^2, Q_0^2) q_i(x_{\alpha}, Q_0^2) \quad (7)$$

where:

$$FK_{j,\alpha} = \sum_i C_i(x, Q^2) \otimes \left[ \sum_{21^{\tau}} \sum_{\beta} \Gamma_{ij,\alpha\beta}^{\tau} I_{\beta}(x) I_{\tau}(Q^2) \right] \quad (8)$$

# Backup

## At LO, DGLAP basis

$$\begin{aligned}F_2^\gamma(x, Q^2, A) &= \frac{1}{A} \left( Z \mathbf{F}_2^p + (A - Z) \mathbf{F}_2^n \right) \\&= \frac{5}{18} \Sigma - \left( \frac{Z}{3A} - \frac{1}{6} \right) T_3 + \frac{1}{18} (T_8 - T_{15}) + \frac{1}{30} (T_{24} - T_{35}) \\&= \frac{2}{9} \Sigma - \left( \frac{Z}{3A} - \frac{1}{9} \right) T_3\end{aligned}$$

## Evolution ( $Q_0 < M_c$ )

$$\Sigma(Q^2) = \Gamma_{qq} \Sigma(Q_0^2) + \Gamma_{qg} g(Q_0^2)$$

$$g(Q^2) = \Gamma_{gq} \Sigma(Q_0^2) + \Gamma_{gg} g(Q_0^2)$$

$$\Gamma^+ = \Gamma_{T_3} = \Gamma_{T_8}$$

$$T_{15}(Q_0^2) = T_{24}(Q_0^2) = T_{35}(Q_0^2) = \Sigma(Q_0^2)$$

**3 independent distributions**

$$\Sigma, g, T_3$$

## Evolution basis

$$\Sigma = \sum_{i=1}^{n_f} q_i^+ \quad \text{where: } q^\pm = q \pm \bar{q}$$

$$T_3 = u^+ - d^+$$

$$T_8 = u^+ + d^+ - 2s^+$$

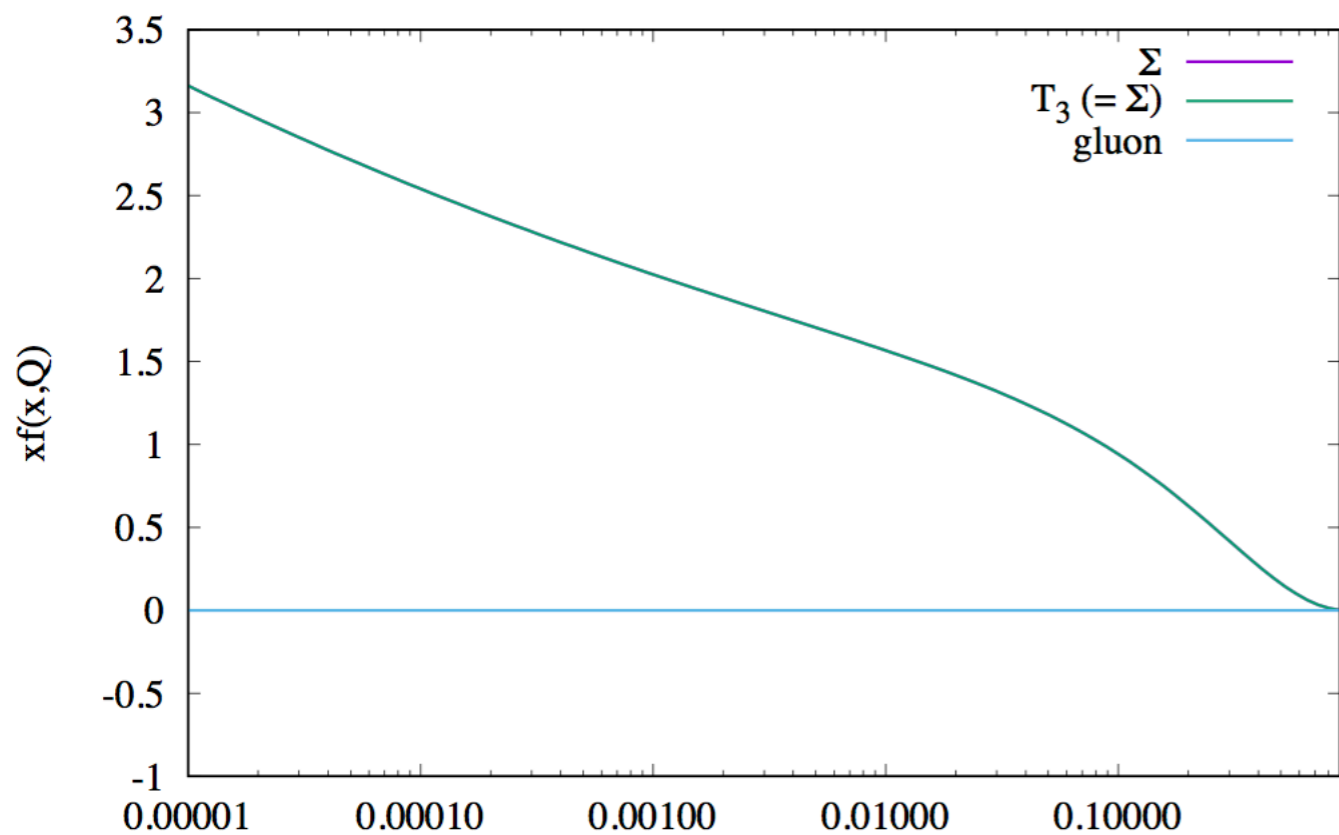
$$T_{15} = u^+ + d^+ + s^+ - 3c^+$$

$$T_{24} = u^+ + d^+ + s^+ + c^+ - 4b^+$$

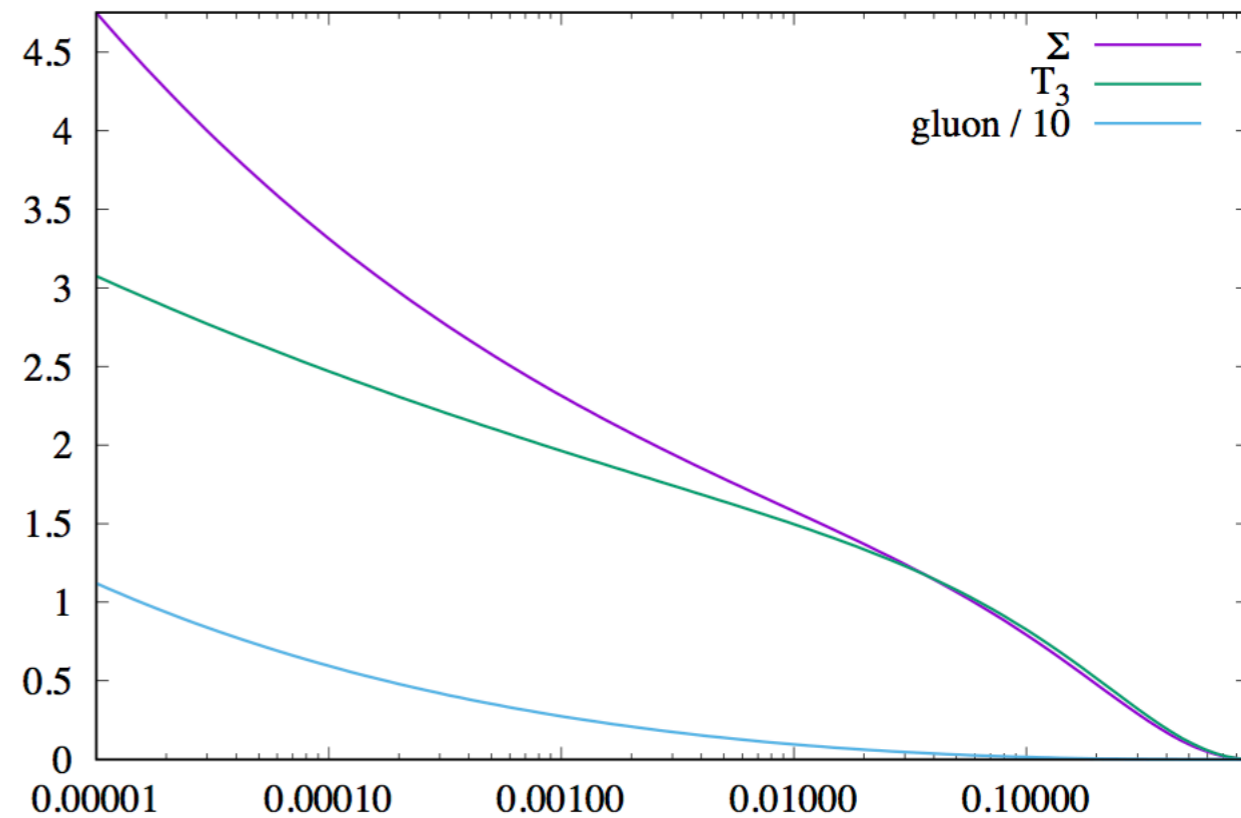
$$T_{35} = u^+ + d^+ + s^+ + c^+ + b^+ - 5t^+$$

# Backup

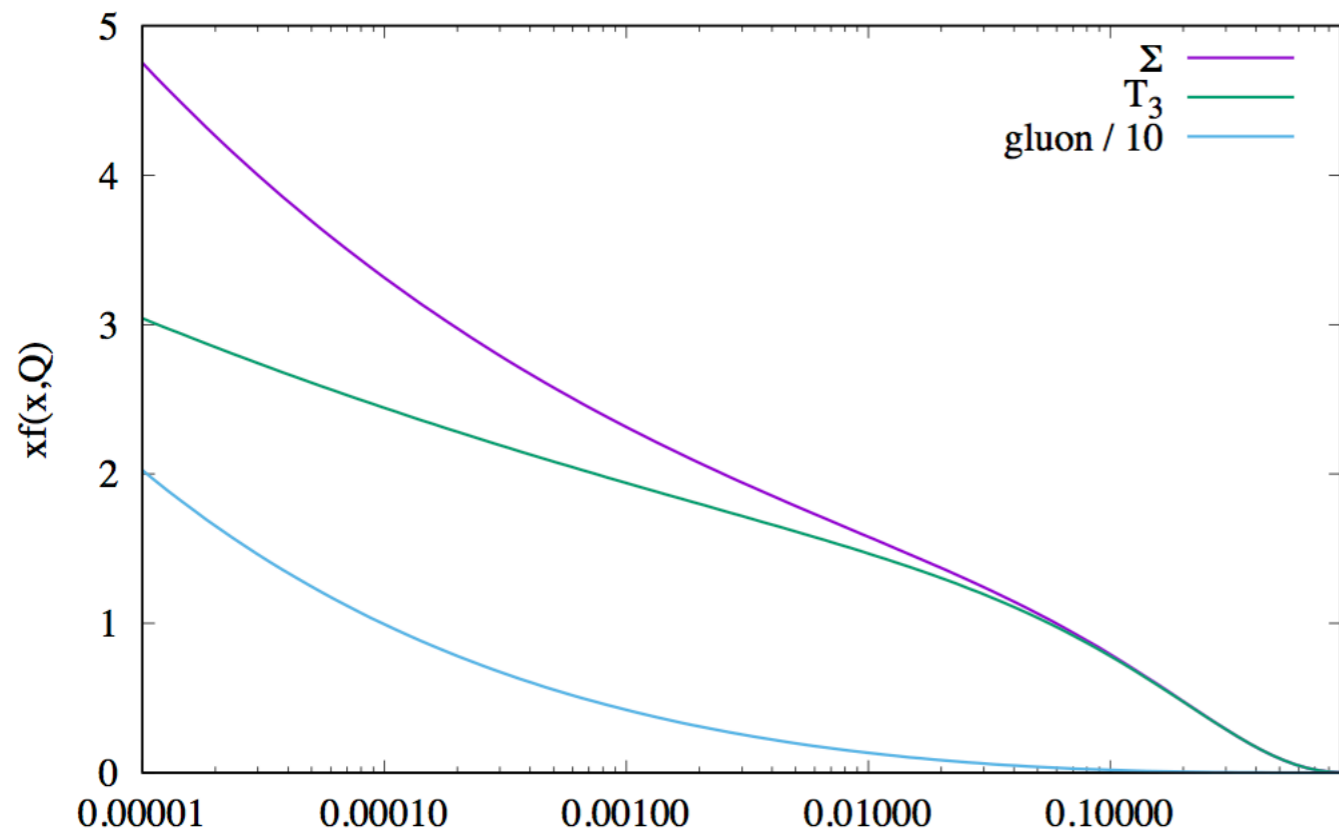
Q = 1.14 GeV (initial scale)



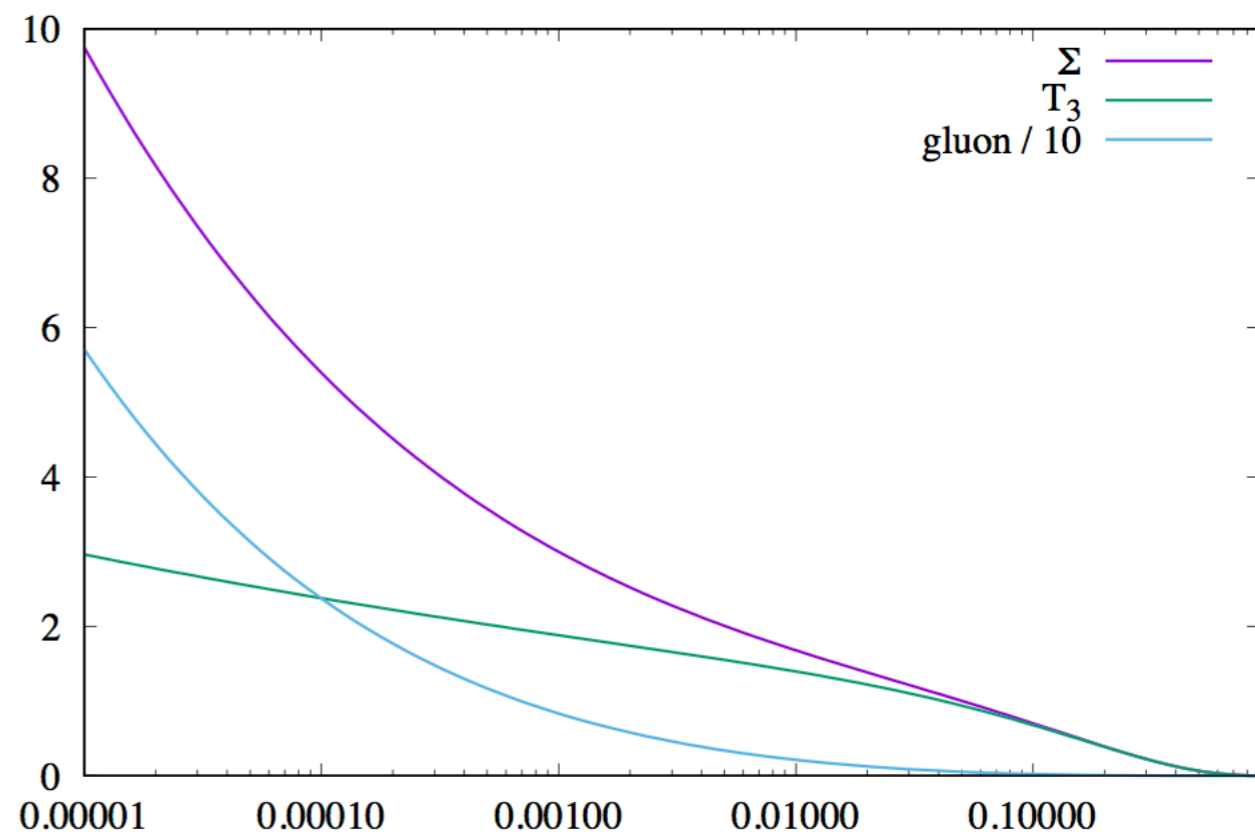
Q = 5 GeV



Q = 10 GeV



Q = 100 GeV



# Results (2) - WarmUp

## Constraints on A-dependance

Testing the assumption  $NN(x,A) = A^n f(x)$  via  $\frac{d \ln(NN(x,A))}{d \ln(A)} = n(x)$

For  $x = 0.121258$

