

A data-driven test of a quantum-statistics PDF parametrisation.

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Consider Deep Inelastic Scattering (DIS)

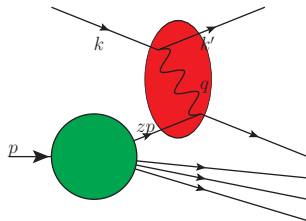
$$\sigma(x_{Bj}, Q^2, \{P\}) = \sum_{i \in \{q, \bar{q}, g\}} \int_{x_{Bj}}^1 \frac{dx}{x} C_i\left(\frac{x_{Bj}}{x}, \alpha_s, Q^2\right) f_i(x, Q^2, \{P\}) \quad (1)$$

To fit $f_i(\dots)$ at all scales:

1. Parametrise f at some (low) initial scale
 $Q_0 \rightarrow f_i(x, Q_0^2, \{P\})$
2. Use DGLAP evolution from Q_0 to Q

$$f_i(x, Q_0^2, \{P\}) \xrightarrow{\text{DGLAP}} f_i(x, Q^2, \{P\})$$

3. Take convolution with parton-level observable (1)
4. "Minimize" the difference
 $\sigma_{\text{exp}}(x_{Bj}, Q^2) - \sigma(x_{Bj}, Q^2, \{P\})$ by varying the free parameters $\{P\}$



No analytical way to pick the initial parametrisation nor Q_0

- HERAPDF $\rightarrow f_i(x, Q_0^2) = x^\alpha(1-x)^\beta P_n(z)$ using polynomials [1506.06042]
- NNPDF $\rightarrow f_i(x, Q_0^2) = x^\alpha(1-x)^\beta \text{NN}(z)$ using neural networks [2109.02653]
- ... many others [2207.04739],[1912.10053]

An alternative to a generic parametrisation is using physical arguments to model the structures of the proton...

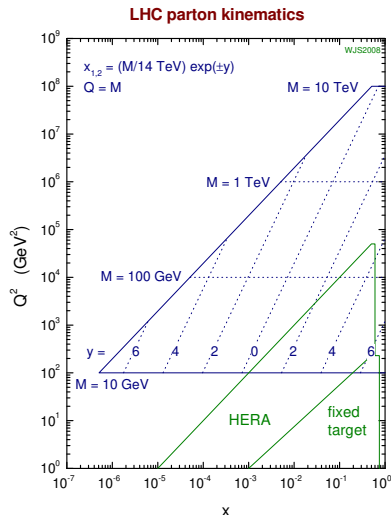
Proton \rightarrow gas mixture of massless partons at equilibrium

[hep-ph/0109160]

- Quarks(gluons) distributions behave according to Fermi(Bose) statistics

$$f_i(x, Q_0^2) \propto \left[\exp\left(\frac{x - X_{0i}}{\bar{x}}\right) \pm 1 \right]^{-1};$$

- “Chemical potentials” of quarks and antiquarks are related $X_{0q}^{\downarrow} = -X_{0\bar{q}}^{\downarrow}$
- Gluons behave like blackbody radiation $\rightarrow X_{0g} = 0$
- Universal “temperature” parameter \bar{x}



We summarize the expressions from the model as

[hep-ph/0109160]

$$h(x; b, \bar{x}, X) = \frac{x^b}{\exp\left(\frac{x-X}{\bar{x}}\right) + 1},$$

$$xf_{q\uparrow}(x, Q_0^2) = AX_q^{\uparrow\downarrow} h(x; b, \bar{x}, X_q^{\uparrow\downarrow}) + \tilde{A}h(x; \tilde{b}, \bar{x}, 0), \quad (2a)$$

$$xf_{\bar{q}\uparrow}(x, Q_0^2) = \bar{A} \frac{1}{X_q^{\uparrow\downarrow}} h(x; \bar{b}, \bar{x}, -X_q^{\uparrow\downarrow}) + \tilde{A}h(x; \tilde{b}, \bar{x}, 0), \quad (2b)$$

with $q \in \{u, d\}$,

$$xf_g(x, Q_0^2) = \frac{A_g x^{b_g}}{\exp(x/\bar{x}) - 1}. \quad (2c)$$

An auxiliary term $\tilde{A}h(x; \tilde{b}, \bar{x}, 0)$ is introduced to control the high-energy region.
 Fitting only unpolarised DIS data \rightarrow average over spin

$$f_q(x, Q_0^2) = f_{q\uparrow}(x, Q_0^2) + f_{q\downarrow}(x, Q_0^2)$$

Writing the unpolarised valence and sea contributions ($q \in \{u, d\}$)

$$\begin{aligned}
 xq_v(x, Q_0^2) &= q(x, Q_0^2) - \bar{q}(x, Q_0^2) \\
 &= A \left[X_q^\uparrow h(x; b, \bar{x}, X_q^\uparrow) + X_q^\downarrow h(x; b, \bar{x}, X_q^\downarrow) \right] \\
 &\quad - \bar{A} \left[\frac{1}{X_q^\downarrow} h(x; \bar{b}, \bar{x}, -X_q^\downarrow) + \frac{1}{X_q^\uparrow} h(x; \bar{b}, \bar{x}, -X_q^\uparrow) \right] \tag{3a}
 \end{aligned}$$

$$\begin{aligned}
 x\bar{q}(x, Q_0^2) &= \bar{A} \left[\frac{1}{X_q^\downarrow} h(x; \bar{b}, \bar{x}, -X_q^\downarrow) + \frac{1}{X_q^\uparrow} h(x; \bar{b}, \bar{x}, -X_q^\uparrow) \right] \\
 &\quad + 2\tilde{A}h(x; \tilde{b}, \bar{x}, 0), \tag{3b}
 \end{aligned}$$

$$xg(x, Q_0^2) = \frac{A_g x^{b_g}}{\exp(x/\bar{x}) - 1} \tag{3c}$$

$$s(x, Q_0^2) = \bar{s}(x; b, \bar{x}, X, Q_0^2) = \frac{f_s}{1 - f_s} \bar{d}(x; b, \bar{x}, X, Q_0^2) \quad \text{with} \quad f_s = 0.4. \tag{3d}$$

There are 13 parameters: $\{\bar{x}, A_g, A, \bar{A}, \tilde{A}, X_u^{\uparrow\downarrow}, X_d^{\uparrow\downarrow}, b, \bar{b}, b_g, \tilde{b}\}$

Additional constraints to apply:

- Valence and momentum sum rules \rightarrow fix normalisations $\{A_g, A, \bar{A}\}$
- $b_g = 1 + \tilde{b}$ (Regge theory)
- $\bar{b} = b$

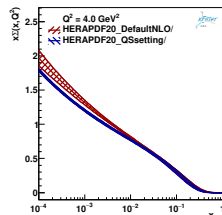
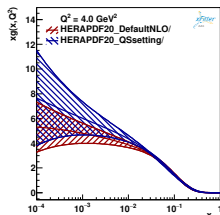
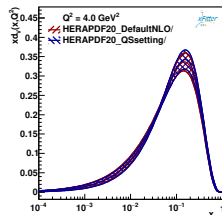
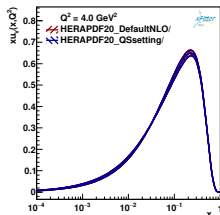
This leaves 8 free parameters to fit: $\{\bar{x}, \tilde{A}, X_u^{\uparrow\downarrow}, X_d^{\uparrow\downarrow}, b, \tilde{b}\}$.

We use `xfitter` public framework with the HERA DIS dataset

[1410.4412].

Some additional constraints are applied:

- Parametrisation scale
 $Q_0^2 = 4 \text{ GeV}^2$
 $\rightarrow \text{cut } Q^2 = 3.5 \text{ GeV}^2 < Q_0^2$
- Theory inputs:
APFEL@NLO
VFNS (FONLL-B)



Benchmark fit between default HERAPDF2.0 NLO configuration in `xfitter` and our settings

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- Theory inputs: \rightarrow
 APFEL@NLO
 VFNS (FONLL-B)
- Improved description of NCep 920 \leftarrow

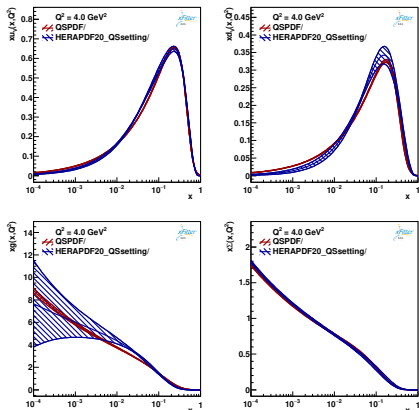
Dataset	HERAPDF20 Default- NLO	HERAPDF20 QSsetting
HERA1+2 CCem	54 / 42	54 / 42
HERA1+2 NCep 820	68 / 70	64 / 68
HERA1+2 NCep 460	217 / 204	216 / 200
HERA1+2 NCep 920	439 / 377	397 / 363
HERA1+2 CCep	43 / 39	45 / 39
HERA1+2 NCem	222 / 159	221 / 159
HERA1+2 NCep 575	219 / 254	217 / 249
Correlated χ^2	86	67
Log penalty χ^2	+8.3	-4.68
Total χ^2 / dof	1357 / 1131	1275 / 1106
χ^2 p-value	0.00	0.00

Benchmark fit between default HERAPDF2.0 NLO configuration in `xfitter` and our settings

Testing the QSPDF parametrisation: fits and χ^2

Now we fit QSPDF against the HERA DIS dataset.

- Reduced error bands
- Only experimental uncertainty is accounted for



In blue the benchmark fit with the HERAPDF2.0 and in red the fit of QSPDF

Now we fit QSPDF against the HERA DIS dataset.

- Fit quality is good
- very minimal set of parametres

Dataset	QSPDF	HERAPDF20 QSsetting
HERA1+2 CCep	59 / 39	45 / 39
HERA1+2 CCem	69 / 42	54 / 42
HERA1+2 NCem	229 / 159	221 / 159
HERA1+2 NCep 820	71 / 68	64 / 68
HERA1+2 NCep 920	468 / 363	397 / 363
HERA1+2 NCep 460	231 / 200	216 / 200
HERA1+2 NCep 575	235 / 249	217 / 249
Correlated χ^2	104	67
Log penalty χ^2	-71.03	-4.68
Total χ^2 / dof	1397 / 1112	1275 / 1106
χ^2 p-value	0.00	0.00

	QSPDF	HERAPDF2.0
# param.	8	14
χ^2 /D.O.F.	1.26	1.15

We recover:

- $X_u^\uparrow > X_d^\downarrow \sim X_u^\downarrow > X_d^\uparrow$
like in [hep-ph/0109160]
- Qualitative agreement a determination of parameters
- No information on polarised PDF (unlike previous work)

Parameter	QSPDF	[hep-ph/0109160]
A	3.04	1.75
\bar{A}	0.12	1.91
A_g	33.52	14.28
\tilde{A}	0.133 ± 0.004	0.083
X_d^\uparrow	0.14 ± 0.02	0.23
X_d^\downarrow	0.284 ± 0.007	0.302
X_u^\uparrow	0.419 ± 0.007	0.461
X_u^\downarrow	0.21 ± 0.02	0.298
$b = \bar{b}$	0.52 ± 0.01	0.41
$\tilde{b} = b_g - 1$	-0.173 ± 0.003	-0.253
\bar{x}	0.092 ± 0.001	0.099

Testing the QSPDF parametrisation: $\bar{d} - \bar{u}$ distribution

We compare the $\bar{d} - \bar{u}$ distributions explicitly
→ Interesting qualitative feature reproduced!

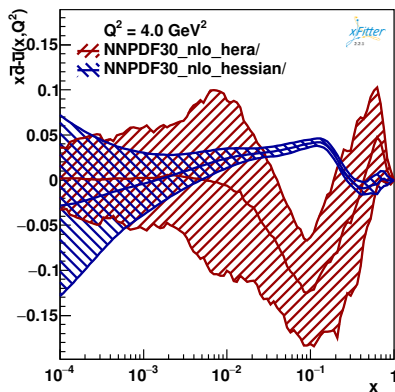


Figure: Comparison of NNP3.0 fits with the HERA dataset only and the default dataset, (NLO theory)

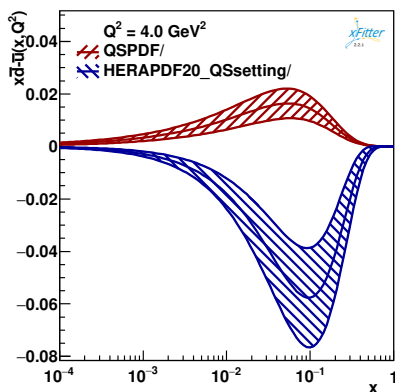


Figure: In blue the benchmark fit with the HERAPDF2.0 parametrisation and in red the fit of QSPDF

Fitting a statistical PDF model

- ✓ We performed a fit a custom PDF parametrisation (QSPDF) against the HERA DIS dataset.
- ↑ Acceptable agreement between data and model.
- ↑ Fit parameters match a previous attempts at fitting a similar parametrisation.
- ↓ This simplest iteration of the parametrisation isn't very competitive against more established models...
- ↑ ... but uses a smaller number of degrees of freedom.
- ↑ Qualitative shape of the $\bar{d} - \bar{u}$ distribution reproduced with HERA data only.

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Outlook

- Improve the current fit.
- Modify the parametrisation (relaxing \bar{b} , add transverse potentials) → building a “minimal” parameter set with state-of-the art performance.

Thank you for your attention

Backup slides

writing the unpolarised valence and sea contributions

$$xq_v^{\uparrow\downarrow}(x; b, \bar{x}, X, Q_0^2) = x[f_{q^{\uparrow\downarrow}}(x; b, \bar{x}, X, Q_0^2) - f_{\bar{q}^{\uparrow\downarrow}}(x; b, \bar{x}, X, Q_0^2)], \quad (4)$$

$$xq_{\text{sea}}^{\uparrow\downarrow}(x; b, \bar{x}, X, Q_0^2) = x\bar{q}^{\uparrow\downarrow}(x; b, \bar{x}, X, Q_0^2) = xf_{\bar{q}^{\uparrow\downarrow}}(x; b, \bar{x}, X, Q_0^2) \quad (5)$$

$$\begin{aligned} xq_v(x; b, \bar{x}, X, Q_0^2) &= x\left[q_v^{\uparrow}(x; b, \bar{x}, X, Q_0^2) + q_v^{\downarrow}(x; b, \bar{x}, X, Q_0^2)\right] = \\ &= A\left[X_q^{\uparrow}f(x; b, X_q^{\uparrow}) + X_q^{\downarrow}f(x; b, X_q^{\downarrow})\right] + \\ &\quad - \bar{A}\left[\frac{1}{X_q^{\downarrow}}f(x; \bar{b}, -X_q^{\downarrow}) + \frac{1}{X_q^{\uparrow}}f(x; \bar{b}, -X_q^{\uparrow})\right] \end{aligned} \quad (6a)$$

$$\begin{aligned} x\bar{q}(x; b, \bar{x}, X, Q_0^2) &= \bar{q}^{\uparrow}(x) + \bar{q}^{\downarrow}(x) = \\ &= \bar{A}\left[\frac{1}{X_q^{\downarrow}}f(x; \bar{b}, -X_q^{\downarrow}) + \frac{1}{X_q^{\uparrow}}f(x; \bar{b}, -X_q^{\uparrow})\right] + 2\tilde{A}h(x), \end{aligned} \quad (6b)$$

$$xf_g(x) = \frac{A_g x^{b_g}}{\exp(x/\bar{x}) - 1}, \quad (6c)$$

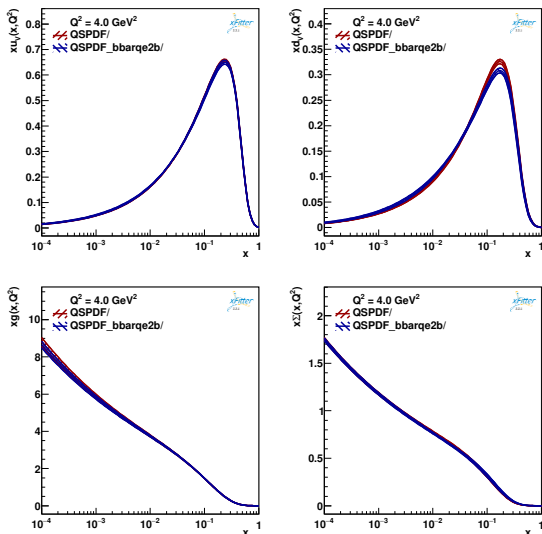


Figure: Effect of different constraint on \bar{b} : $\bar{b} = b$ and $\bar{b} = 2b$

Gluon PDF error in default HERAPDF2.0 fit @NLO

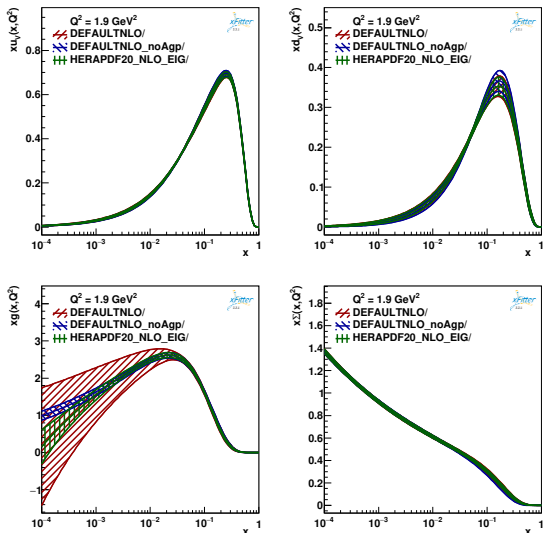


Figure: Large error in the gluon pdf induced by poor determination of $A_{gg} = 0.23 \pm 0.29$.