Discriminating models of meson structure using global QCD fits

Brief remarks for the workshop discussion Aurore Courtoy,¹ Tim Hobbs,² Fred Olness,² <u>Pavel Nadolsky</u>² ¹UNAM, Mexico ²SMU, USA

Based on studies with CTEQ-TEA, nCTEQ, and xFitter collaborations

Key lesson from 35+ years of global fits of nucleon and nuclear PDFs:

Complete error analysis, including all sources of uncertainties, is critical for falsifying theoretical models using experimental data

My apologies for incomplete references!



Are any of these pion models ruled out? Not so fast...

- Nambu and Jona-Lasinio Model:
 - R. Davidson, E. Arriola, PLB (1995)
 - J.T. Londergan *et al.* PLB (1994).
 - T. Shigetani *et al.* PLB (1993).
- Dyson Schwinger Equation:
 - M. Hecht *et al.* PRD (2001).
- Chiral Quark Model:
 - K. Suzuki, W. Weise, NPA (1998).
 - D. Arndt, M. Savage, nucl-th (2001)
- Light-front constituent quark models:
 - Gerry Miller, et al. (too many to list).
- Instanton Model:
 - A. Dorokhov, L. Tomio, PRD (2000)
- QCD Sum Rule Calculations
 - A. Bakulev *et al.* PLB (2001).
- Lattice Gauge
 - C. Best *et al.* PRD (1997).

Paul E. Reimer, 3rd International Workshop on Nucleon Structure at Large Bjorken x



At some base q_0 NJL: $xq(x)/(1-x)^{\beta}\beta = 1$ $pQCD: xq(x)/(1-x)^{\beta}\beta = 2$ DSE: $xq(x)/(1-x)^{\beta}\beta \frac{1}{4}1.9$ Evolution to experimental Q increases β .

Comparing theoretical predictions to phenomenological PDFs

- When fitting a nonperturbative model to cross section data is impractical, the model may be compared to the MS PDFs from a phenomenological set (CTEQ, JAM, NNPDF, etc.) constrained by the global QCD data
- The PDF is a theoretical construct, not a physical observable.
- For example, an unpolarized quark PDF in a hadron $A = p, n, \pi, K, ...$ in the \overline{MS} factorization scheme is defined to all orders of α_s by

 $f_{q/A}(\xi,\mu) = \frac{1}{4\pi} \int_{-\infty}^{+\infty} dy^{-} e^{-i\xi p^{+}y^{-}} \langle A | \bar{\psi}_{q} (0^{+}, y^{-}, \vec{0}_{T}) \gamma^{+} W(A; y^{-}, 0) \psi_{q}(0^{+}, 0^{-}, \vec{0}_{T} | A \rangle$

QCD cross sections factorize in terms of PDFs *f_{a\A}(ξ, μ)* and hard cross sections *ô*. In *Z* production at the LHC:



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T.-J. Hou et al., arXiv:1912.10053

- Modern PDFs such as the CT18 (N)NLO PDF set are provided in the form of central fits and error sets for estimating 68% or 90% C.L. uncertainties
- Functional forms of some PDF sets at the initial scale Q₀ that may be motivated by nonperturbative models. Other sets use "unbiased" ultra-flexible functional forms given by neural networks.
- Mimicry a fundamental feature of multivariate optimization:
 Diverse functional forms of PDFs at the initial scale μ₀ provide equally good description of QCD data from DIS, Drell-Yan pair, jet, and tt



Sources of the uncertainty on PDFs

- **1. Experimental uncertainties**, e.g., statistical, correlated and uncorrelated systematic uncertainties of each experimental data set;
 - can be estimated using the $\Delta \chi^2 = 1$ criterion when experimental sets agree
- 2. Theoretical uncertainties due to the absent higher-order and powersuppressed radiative contributions, QCD scale choices, and uncertainties from using parton showering programs in experimental analyses
- **3. Methodological uncertainties** associated with the selection of experimental data sets, fitting procedures, and goodness-of-fit criteria.
- **4. Parameterization uncertainties** associated with the choice of the PDF functional form;
 - contribute at least a half of the CT18 total PDF uncertainty
 - Rapid variations in $f_{a/A}(\xi, \mu)$ are generally not constrained well

The uncertainty of published CT18 PDFs estimates the sum of four contributions Kovarik et al., arXiv: <u>1905.06957</u>

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"Error sets" for computing PDF uncertainties

- 1. Based on diagonalization of the Hessian matrix (CTEQ, MMHT, ABM, HERAPDF, ...)
 - PDFs at the initial scale Q_0 are given by polynomial functional forms,

 $f(\xi,Q_0) = A_0 \, x^{A_1} (1-x)^{A_2} P(x;A_3,A_4 \dots)$

- 2. Based on Monte-Carlo sampling of probability and neural networks (NNPDF)
 - The $x^{A_1}(1-x)^{A_2}$ dependence is modulated by a neural network function $P_{NN}(x)$ that can oscillate
 - A nonpert. model can predict such replicas in an x-averaged sense

See, e.g., Nocera, 1410.7290; Ball, Nocera, Rojo, 1604.00024



10⁻³

10-2

10-1

10-4

10-5

Example: the effective $(1 - x)^{A_2}$ dependence of nucleon and pion PDFs Drell, N

Counting rules for a quark distribution with $x \to 1$ in a hadron A moving with $p^+ \gg 1 \text{ GeV}$, with dependence on helicities: $\lim_{x \to 1} f(x, \mu_0 \sim 1 \text{ GeV}, \lambda_q) \propto (1-x)^{2 n_s - 1 + 2|\lambda_q - \lambda_A|}$

 n_s is the number of spectator fermions, λ_q and λ_A are helicities of struck quark and hadron A

Nucleon valence PDFs: $n_s = 2$, $\lambda_A = \pm \frac{1}{2}$, $|\lambda_q - \lambda_A| = 0$ or 1; $\lim_{x \to 1} \sum_{\lambda_q, \lambda_A} f_{q/p}(x, \mu_0, \lambda_q) \propto (1 - x)^{A_2} \text{ with } A_2 = 3$

Pion valence PDFs: $n_s = 1$, $\lambda_A = 0$, $|\lambda_q - \lambda_A| = 1$; $\lim_{x \to 1} \sum_{\lambda_q, \lambda_A} f_{q/\pi}(x, \mu_0, \lambda_q) \propto (1 - x)^{A_2} \text{ with } A_2 = 2$

- A_2 is the same for u and d quarks
- At $\mu > \mu_0$, the effective A_2 increases because of DGLAP evolution

 $n_{z} = 2$

Drell, Yan, West, PRL 24 (1970) 1206 Ezawa, Nuovo Cim., A23 (1974) 271 Farrar, Jackson, PRL, 35 (1975) 1416 Berger, Brodsky, PRL 42, 940 Brodsky, Lepage, 1979 and many others

Do phenomenological PDFs agree with Brodsky-Farrar quark counting rules?

Note: the PDF parametrizations are complex. See CT18 parametrizations in the backup. We will examine 363 CT18 candidate parametrization forms.

If at
$$x \to 1$$
:

$$f(x, Q_0) = \underbrace{(1 - x)^{A_2^{true}}}_{fast function} \times \underbrace{\Phi(x)}_{slow function},$$
then
$$A_{2,eff}(x) \equiv \frac{\partial \ln(f(x, Q_0))}{\partial \ln(1 - x)} \approx A_2^{true} + small term.$$

form for u_v, d_v

$$q_v(x, Q = Q_0) = x^{a_1 - 1} (1 - x)^{a_2} P_a^v(y) \qquad y \equiv \sqrt{x}$$

$$P_a^v(y) = a_3(1-y)^4 + a_44y(1-y)^3 + a_56y^2(1-y)^2 + a_64y^3(1-y) + y^4$$
$$a_6 = 1 + \frac{1}{2}a_1$$

more flexible form (i.e., high-order polynomial) used for SU(2) sea PDFs;
 e.g.:

$$\bar{d}(x, Q = Q_0) = x^{a_1 - 1} (1 - x)^{a_2} P_a^d(y)$$

$$P_a^{\bar{d}}(y) = a_3 (1 - y)^5 + a_4 5y (1 - y)^4 + a_5 10y^2 (1 - y)^3$$

$$+ a_6 10y^3 (1 - y)^2 + a_7 5y^4 (1 - y) + y^5$$

Effective $(1-x)^{A_2}$ 4.0 dependence, proton $\frac{3.5}{5}$

 $A_{2,eff}(x) \equiv \frac{\partial \ln(f(x,Q_0))}{\partial \ln(1-x)}$ is computed at Q = 1.3, 2, 10GeV; and x=0.85, 0.5, 0.3.

The scatter points are for 363 distinct parametric forms of CT18 NNLO candidates.

3679 data points in the fit, variation in χ^2 is within 100 units for the shown fits

The fitted data are below x = 0.75. Extrapolation to x = 0.85.





Nadolsky, Olness, Courtoy, Hobbs, Pion-Kaon Workshop @ JLab

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CT18 NNLO candidates, effective (1-x) powers for valence guarks in the nucleon

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Does this observation rule out the $(1 - x)^2$ dependence?

Effective $(1-x)^{A_2}$ dependence, **pion**

Very similar PDFs can be obtained using distinct functional forms



FIG. 2. The valence distribution when using minimal parameterisation $(D_v = 0)$ and the extended parameterisation with free D_v . The shown uncertainty bands do not include scale variations. The high-x behavior is linear in (1 - x).

$$\lim_{x \to 1} \frac{v_1(x)}{v_2(x)} = 21.6 \ (1-x)^{0.69} + \cdots$$

Mimicry I

xFitter Pion NLO PDFs, arXiv:2002.02902



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Effective $(1 - x)^{A_2}$ dependence, **pion**



Mimicry II

Bednar, Cloët, Tandy, 2018

Predict pion PDFs from a truncated DSE approach at low scale Q = 0.78 GeV, $xv(x) = A_0(1-x)^2 \times [fancy polynomial] + c_4(1-x)^4$

The form at Q=0.78 GeV is quantitatively close to

 $xv(x) = 1.74 (1 - x)^{0.88}$; evolves to xv(x) that is nearly linear in (1 - x) in the DY data region, agrees with JAM PDFs



Pion PDFs at $\zeta_5 = 5.2 \text{ GeV}$



ⓒ M. Ding shows that pion DY data are compatible with the form $q_π(x) = A_0 x^{-0.153} (1 - x)^{2+ε} × [fancy polynomial]$

 \otimes It is often unclear how well individual models agree with the data. Do they have good χ^2 , are there systematic disagreements in some kinematic regions?

⇒ Good fits must pass multiple statistical tests, see Kovarik, PN, Soper, <u>1905.06957</u>

QCD scale dependence in the Drell-Yan process



Instructive to repeat the NLL resummation calculation by Aicher, Shäfer, Vogelsang (2010) at NNLO with complete studies of uncertainties due to matching, nonpert. contributions, nuclear PDFs, and correlation with overall normalization

Factorization scale dependence of Drell-Yan cross sections is generally mild at NLO, but grows toward largest x values.

At $x \ge 0.5$, NLL threshold resummation:

- reduces the scale dependence
- increases A_2 from 1 at NLO with $\mu = M_{\ell \bar{\ell}}$ to 2.34
- no large $\ln(1-x)$ terms evident at smaller x



Threshold resummation for DY process

Aicher, Shafer, Vogelsang, arXiv:1009.2481, and references therein

$$\begin{aligned} \text{When } z \equiv \frac{Q^2}{\hat{s}} \to 1; \qquad \frac{\mathrm{d}\sigma}{\mathrm{d}Q^2\mathrm{d}\eta} &= \int_{-\infty}^{\infty} \frac{\mathrm{d}M}{2\pi} e^{-iM\eta} \int_{C-i\infty}^{C+i\infty} \frac{\mathrm{d}N}{2\pi i} \tau^{-N} \sigma(N, M). \equiv W \\ & \ln \tilde{\omega}_{q\bar{q}} = C_q \left(\frac{Q^2}{\mu^2}, \alpha_s(\mu^2)\right) + 2 \int_0^1 \mathrm{d}\zeta \frac{\zeta^{N-1} - 1}{1 - \zeta} \\ \sigma(N, M) &= \sigma_0 \sum_{a,b} f_a^{\pi, N+i\frac{M}{2}} f_b^{A, N-i\frac{M}{2}} \tilde{\omega}_{ab}(N, M). \qquad \times \int_{\mu^2}^{(1-\zeta)^2 Q^2} \frac{\mathrm{d}k_\perp^2}{k_\perp^2} A_q(\alpha_s(k_\perp)), \end{aligned}$$

where $\tilde{\omega}_{ab}$ resum the soft and collinear gluon contributions:

The resummed *W* term dominates when log(1 - z) terms are large. In the kinematics of pion DY data, typical $z \ll 1$, the full description requires matching the *W* term on the fixed-order (**FO**) cross section at (N)NLO in α_s :

All z:
$$\frac{d^2\sigma}{dQ^2d\eta} = W + FO - overlap$$

Final remarks

- 1. A theoretical model cannot be falsified empirically if its uncertainty is unknown.
- 2. Multiple functional forms of PDFs adequately describe the global set of experimental data.
- 3. The *x* –dependence of the best-fit PDF depends on the PDF definition (e.g., \overline{MS} factorization scheme), order of α_s , treatment of power-suppressed corrections, and other factors. Comparison to the pheno PDF requires proper conversion into the definition adopted by the pheno PDF.
- 4. Threshold resummation may modify both the shapes and normalizations of some pion DY cross sections. The resummed W term dominant at $Q^2/\hat{s} \rightarrow 1$ must be matched to the fixed-order (FO) cross section at Q^2/\hat{s} < 1. The resulting PDFs may depend on the resummed nonpertubative contributions at large Mellin moments N and the matching procedure.
- 5. Determination of sea quark and gluon PDFs in the pion generally benefits from the extended reach in Q in DIS and other processes.

Thank you for your attention

form for u_v, d_v

$$q_v(x, Q = Q_0) = x^{a_1 - 1} (1 - x)^{a_2} P_a^v(y) \qquad y \equiv \sqrt{x}$$

$$P_a^v(y) = a_3(1-y)^4 + a_4 4y(1-y)^3 + a_5 6y^2(1-y)^2 + a_6 4y^3(1-y) + y^4$$
$$a_6 = 1 + \frac{1}{2}a_1$$

more flexible form (i.e., high-order polynomial) used for SU(2) sea PDFs;
 e.g.:

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$$+ a_6 10y^3 (1 - y)^2 + a_7 5y^4 (1 - y) + y^5$$

form for u_v, d_v

$q_v(x, Q = Q_0) = x^{a_1 - 1} (1 - x)^{a_2} P_a^v(y)$ $y \equiv \sqrt{x}$

cent. fitted parameter, CT18	u_v	d_v	g	\bar{u}	d	s
a_1	0.7632	0.7632	0.5310	-0.0219	-0.0219	-0.0219
a_2	3.0361	3.0361	3.1481	7.7366	7.7366	10.3099
a_3	1.5019	2.6141	3.0314	(4)	(4)	(4)
a_4	-0.1467	1.8275	-1.7049	0.6179	0.2922	0.4660
a_5	1.6711	2.7203		0.1949	0.6470	0.4660
a_6				0.8709	0.4749	0.2253
a_7				0.2667	0.7414	0.2253
a_8				0.7332		(1)
cent. fitted parameter, $\mathbf{CT18Z}$	u_v	d_v	g	$ar{u}$	d	s
a_1	0.7867	0.7867	0.2892	0.0096	0.0096	0.0096
a_2	3.1480	3.1480	1.8720	8.2730	8.2730	11.3771
a_3	1.5586	3.5016	3.5377	(4)	(4)	(4)
a_4	-0.0750	1.8654	-1.6654	0.6791	0.2999	0.6531
a_5	1.6052	3.5986		-0.0016	0.5322	0.6531
a_6				1.0851	0.7533	0.0539
a_7				0.0455	0.4404	0.0539
a_8				0.7592		(1)

 $P_a^d(y) = a_3(1-y)^5 + a_45y(1-y)^4 + a_510y^2(1-y)^3$

 $+a_6 10y^3(1-y)^2 + a_7 5y^4(1-y) + y^5$

Explore various non-perturbative parametrization forms of PDFs



- CT18par a sample of **some** non-perturbative parametrization forms tried in CT18
- No data constrain very large *x* or very small *x* regions.