

Parton Distribution Functions (Part Deux)

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Introduction: Parton Distribution Functions

<http://www.math.columbia.edu/~woit/wordpress/?p=3715>

“String theory has always been much more popular than phenomenology. Given the choice of **working on the theory that unifies all physics and explains the big bang** or **computing grungy things about parton distribution functions**, it’s easy to see why this is.” [Peter Woit, comment on his blog “*Not Even Wrong*”, 19th May 2011]

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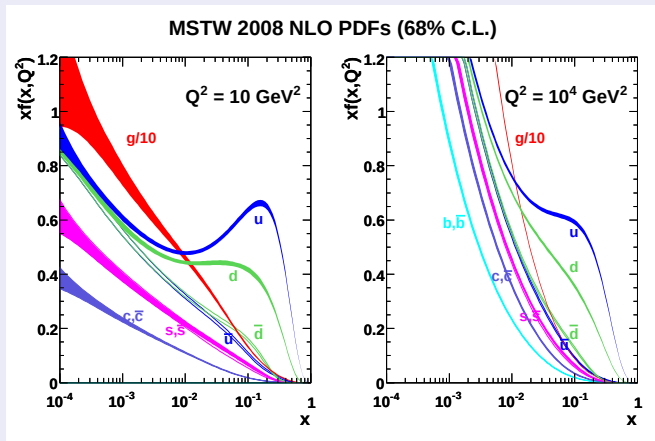
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This talk:

Computing ~~grungy~~ **necessary and interesting** things about **Parton Distribution Functions**.

MSTW 2008 PDFs [<http://projects.hepforge.org/mstwpdf/>]

A. D. Martin, W. J. Stirling, R. S. Thorne, G. Watt [[arXiv:0901.0002](https://arxiv.org/abs/0901.0002)]



- **Error bands** shown are obtained from propagation of **experimental** uncertainties on the fitted data points.

Benchmark comparisons of PDFs from different groups

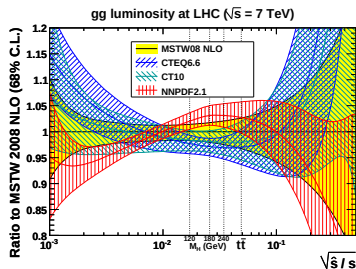
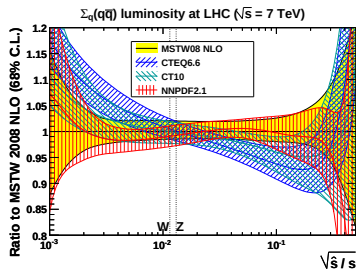
- Various fitting groups currently produce PDF sets:
MSTW, **CTEQ-TEA**, **NNPDF**, **HERAPDF**, **ABKM**, **(G)JR**.
- Results often do not agree within the quoted uncertainties.
- Quantifying and understanding differences *between* groups is as (if not *more*) important as continued improvements *within* groups.
- Systematic comparisons performed over last couple of years
[G.W., JHEP **09** (2011) 069, arXiv:1106.5788].
- Apparent major differences from classes of data fitted:

	MSTW08	CT10	NNPDF2.1	HERAPDF1.5	ABKM09	JR09
HERA DIS	✓	✓	✓	✓	✓	✓
Fixed-target DIS	✓	✓	✓	✗	✓	✓
Fixed-target DY	✓	✓	✓	✗	✓	✓
Tevatron W,Z	✓	✓	✓	✗	✗	✗
Tevatron jets	✓	✓	✓	✗	✗	✗

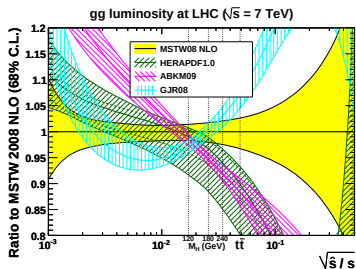
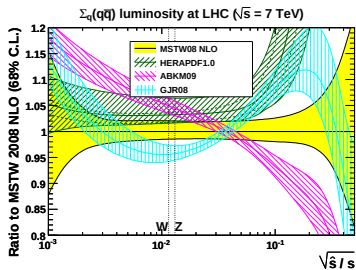
- “Global” $\equiv$ includes all five main categories of data.
- Next generation of global fits will also include **LHC** data.

Partonic luminosities [G.W., JHEP 09 (2011) 069, arXiv:1106.5788]

MSTW vs. other global fits



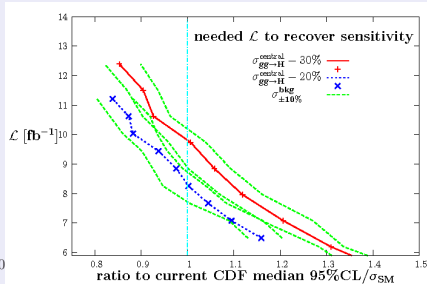
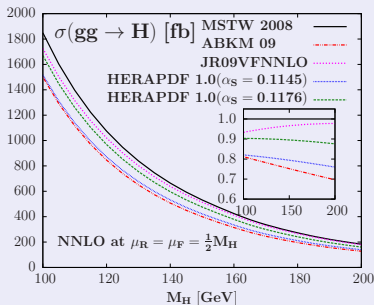
MSTW vs. non-global fits



Tevatron Higgs exclusion limits: a critical appraisal

- $\sigma(gg \rightarrow H)$ depends on gluon distribution at $x \sim M_H/\sqrt{s}$.
- Tevatron and LHC exclusion based on **MSTW 2008** PDFs.
- **PDF**+ α_S uncertainty at 90% C.L. is $\approx 10\%$ at Tevatron.
- But some **non-global** PDF sets can give much *lower* $\sigma(gg \rightarrow H)$.

[J. Baglio, A. Djouadi, S. Ferrag, R. M. Godbole, arXiv:1101.1832]

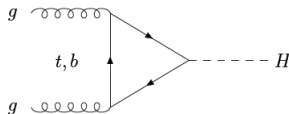
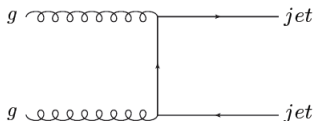


Jets as a discriminator of the high- x gluon distribution

JETS

PV: Any PDF should reproduce jet data if being used for Higgs

Closest observable to Higgs in terms of Luminosity, kinematics and power of coupling!



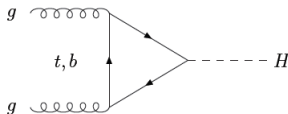
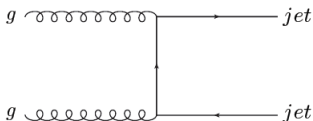
[D. de Florian, talk at "Higgs Hunting 2011", Orsay, France, 28th July 2011]

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- Do PDFs from non-global fits describe Tevatron jet data?
[R. S. Thorne and G.W., JHEP **08** (2011) 100, arXiv:1106.5789]
- Values of $\chi^2/N_{\text{pts.}}$ for CDF Run II inclusive jet data:

NNLO PDF	$\alpha_S(M_Z^2)$	$\mu = p_T/2$	$\mu = p_T$	$\mu = 2p_T$
MSTW08	0.1171	1.39	0.69	0.97
ABKM09	0.1135	2.55	2.76	3.41

Experimental error propagation in global PDF fits

Hessian method (used by MSTW and CTEQ groups)

- Based on *covariance* (inverse *Hessian*) matrix of ≈ 20 fit parameters: diagonalise to produce *eigenvector* PDF sets.
- Varying values of tolerance $T = \sqrt{\Delta\chi^2}$ used to accommodate minor data inconsistencies (average $T \approx 3$ in MSTW08 fit).

Monte Carlo sampling using data replicas (used by NNPDF)

- Generate replica data sets with shifted central values:

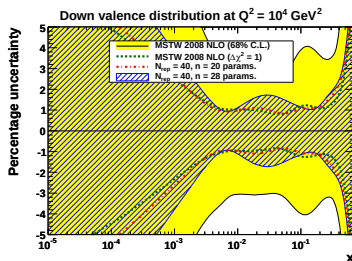
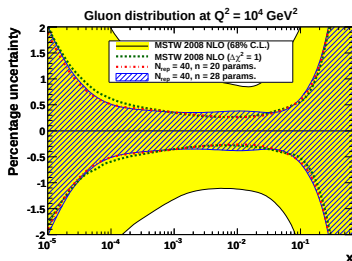
$$D_{m,i} \rightarrow \left(D_{m,i} + R_{m,i}^{\text{uncorr.}} \sigma_{m,i}^{\text{uncorr.}} + \sum_{k=1}^{N_{\text{corr.}}} R_{m,k}^{\text{corr.}} \sigma_{m,k,i}^{\text{corr.}} \right) \cdot (1 + R_m^{\mathcal{N}} \sigma_m^{\mathcal{N}})$$

- Calculate average and s.d. over $N_{\text{rep}} \sim \mathcal{O}(100)$ PDF sets.
- Equivalent to Hessian method with tolerance $T = 1$.
- But more useful when fitting weakly-constrained parameters.

Puzzle: NNPDF *w/o* tolerance $\approx$ MSTW/CTEQ *with* tolerance.

Hessian versus Monte Carlo sampling in MSTW fit

- Compare Hessian **with** or **without** tolerance (20 parameters) to Monte Carlo sampling (either **20** or **28** parameters).

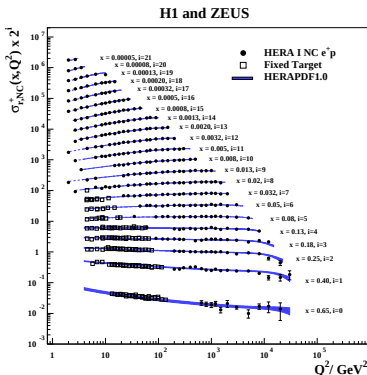


- MC sampling with 20 parameters $\approx$ Hessian with $\Delta\chi^2 = 1$.
- Moving from 20 $\rightarrow$ 28 parameters using MC sampling gives indication of parameterisation bias, i.e. moving from 4 $\rightarrow$ 7 parameters for input xg and 3 $\rightarrow$ 4 parameters for input xd_v .
- Conclusion:** parameterisation bias is likely to be small.

Summary of recent work

- Detailed **PDF comparisons** of benchmark LHC cross sections
⇒ largest differences from global versus non-global fits
[G.W., JHEP **09** (2011) 069, [arXiv:1106.5788](#)].
- PDF dependence of **Higgs cross sections** at Tevatron and LHC
⇒ importance of Tevatron jets to pin down high- x gluon
[R. S. Thorne and G.W., JHEP **08** (2011) 100, [arXiv:1106.5789](#)].
- *Monte Carlo sampling* in context of **MSTW** fit
⇒ **parameterisation bias** is likely to be small [[work in progress](#)].
- **Future:** making use of **LHC data** to constrain PDFs.

Scaling violations of DIS structure functions



$$F_2^{\text{LO}}(x, Q^2) = \sum_q e_q^2 \times f_{q/p}(x, Q^2)$$

Differentiate and insert DGLAP:

$$\Rightarrow \frac{\partial F_2^{\text{LO}}(x, Q^2)}{\partial \ln Q^2} = \frac{\alpha_S}{2\pi} \left[P_{q \leftarrow q}^{\text{LO}} \otimes F_2^{\text{LO}} + \sum_q e_q^2 P_{q \leftarrow g}^{\text{LO}} \otimes f_{g/p} \right]$$

- F_2 falls with Q^2 at large x and rises with Q^2 at small x .
- $\Rightarrow$ Measure *scaling violations* to constrain **small- x gluon**.
- **But** no direct constraint on **large- x gluon** from inclusive DIS: only momentum sum rule and *imposed* parameterisation constraint (non-global fits typically only have 2 or 3 gluon parameters).

Input parameterisation in MSTW 2008 NLO fit

Input parameterisation ($Q_0^2 = 1 \text{ GeV}^2$) in MSTW 2008 fit

$$xu_v = A_u x^{\eta_1} (1-x)^{\eta_2} (1 + \epsilon_u \sqrt{x} + \gamma_u x)$$

$$xd_v = A_d x^{\eta_3} (1-x)^{\eta_4} (1 + \epsilon_d \sqrt{x} + \gamma_d x)$$

$$xS = A_S x^{\delta_S} (1-x)^{\eta_S} (1 + \epsilon_S \sqrt{x} + \gamma_S x)$$

$$x(\bar{d} - \bar{u}) = A_\Delta x^{\eta_\Delta} (1-x)^{\eta_S+2} (1 + \gamma_\Delta x + \delta_\Delta x^2)$$

$$xg = A_g x^{\delta_g} (1-x)^{\eta_g} (1 + \epsilon_g \sqrt{x} + \gamma_g x) + A_{g'} x^{\delta_{g'}} (1-x)^{\eta_{g'}}$$

$$x(s + \bar{s}) = A_+ x^{\delta_S} (1-x)^{\eta_+} (1 + \epsilon_S \sqrt{x} + \gamma_S x)$$

$$x(s - \bar{s}) = A_- x^{0.2} (1-x)^{\eta_-} (1 - x/x_0)$$

- A_u , A_d , A_g and x_0 are determined from sum rules.
- 28 parameters allowed to go free to find overall best fit.
But restrict to 20 parameters for Hessian error propagation.
- MC sampling allows all 28 parameters free $\Rightarrow$ check potential bias.