PDFs and top physics

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[A. Djouadi and S. Ferrag, hep-ph/0310209]



 Errors on PDFs are in some cases the dominating theoretical error on precision observables

Ex.
$$\sigma(Z^0)$$
 at the LHC: $\delta_{PDF} \sim 3\%$, $\delta_{NNLO} \sim 2\%$

[J. Campbell, J. Huston and J. Stirling, (2007)]



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Errors on PDFs might reduce sensitivity to New Physics





PDF determination

Recent and ongoing effort

- Considerable effort and progress in understanding features of PDF fits in recent years
 - HERA-LHC Workshop (2004-2007)
 - PDF4LHC (2007-ongoing)
 - Systematic bemchmarking of predictions for Standard Candle processes from different PDF sets
 - LHAPDF as a common interface to easily access up-to-date PDF fits
- Discussion points:
 - Global vs. Restricted dataset fits
 - Parametrization bias
 - Heavy Flavour contributions treatment
 - Error determination
 - Handling of inconsistencies among datasets
 - Combination of PDF and α_s uncertainties





PDF fits CTEQ-TEA

[Q.-H. Cao, J. Houston, H.-L. Lai, P. Nadolsky, J. Pumplin, D. Stump, W. K. Tung and C.-P. Yuan]

- Latest release: CTEQ6.6
- Dataset: Global (DIS, Drell-Yan Inclusive Jet at Tevatron)
- Perturbative order: NLO (K factors for hadronic observables)
- Heavy Flavours: GM-VFNS (ACOT- χ)
- α_s: Fixed in the fit (α_s(M_Z) = 0.118), sets with different α_s values avbailable
- Parametrization: Standard functional form (22 parameters)
- Error treatment: Hessian Tolerance ($\Delta \chi^2 = 100$)

[A. Martin, J. Stirling, R. Thorne and G. Watt]

- Latest release: MSTW2008
- Dataset: Global (DIS, Drell-Yan, Inclusive Jet at Tevatron)
- Perturbative order: LO/NLO/NNLO (K factors for Drell-Yan)
- Heavy Flavours: GM-VFNS (TR)
- α_s: Determined in the fit alongside PDF parameters
- Parametrization: Standard functional form (29 (20) parameters)
- Error treatment: Hessian Tolerance ($\langle \Delta \chi^2 \rangle \sim 25$)





[R. D. Ball, L. Del Debbio, S. Forte, J. I. Latorre, J. Rojo, M. Ubiali and AG]

- Latest release: NNPDF2.0
- Dataset: Global (DIS, Drell-Yan, Inclusive Jet at Tevatron)
- Perturbative order: NLO (Exact)
- Heavy Flavours: ZM-VFNS
- α_s: Fixed in the fit (α_s(M_Z) = 0.119), sets with different α_s values avbailable
- Parametrization: Neural Networks (259 parameters)
- Error treatment: Monte Carlo



[S. Alekhin, J. Blümlein, S. Klein and S. O. Moch]

- Latest release: ABKM09
- Dataset: DIS and Fixed Target Drell-Yan
- Perturbative order: NLO/NNLO (Exact)
- Heavy Flavours: 3F-/5F-FFNS
- α_s : Determined in the fit alongside PDFs parameters
- Parametrization: Standard functional form (24 parameters)
- Error treatment: CME ($\Delta \chi^2 = 1$)



[H1 & Zeus Collaborations]

- Latest release: HERAPDF1.0
- Dataset: Only combined HERA-I data
- Perturbative order: NLO (Exact)
- Heavy Flavours: GM-VFNS (TR)
- α_s: Fixed in the fit (α_s(M_Z) = 0.1176), sets with different α_s values available
- **Parametrization**: Standard functional form (11 parameters) + model variations
- Error treatment: Hessian/Offset ($\Delta \chi^2 = 1$)

[M. Glück, P. Jimenez-Delgado and E. Reya]

- Latest release: GJR08/JR09
- Dataset: DIS, Fixed Target Drell-Yan
- Perturbative order: LO/NLO/NNLO (Exact)
- Heavy Flavours: FFNS/VFNS
- *α_s*: Determined in the fit
- **Parametrization**: Standard functional form (13 parameters), dynamical PDFs assumption
- Error treatment: Hessian ($\Delta \chi^2 = 1$)



tt cross-section

PDF sets comparison

[G. Watt, PDF4LHC Workshop]



Global fits predictions agree within 1 – σ if combined α_s + PDF uncertainties are taken into account.



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PDF and tops

tt cross-section

PDF sets comparison

• The *t*t cross-section probes the *gg* luminosity



tt cross-section

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PDF Uncertainties for LHC analyses

PDF4LHC recomendation for Higgs Cross Section

At NLO

- Use at least the prediction of the 3 global fits: CTEQ, MSTW and NNPDF.
- Inclusion of other sets (ABKM, HERAPDF and GJR) is optional.
- Use the **envelope** provided by central values and **PDF**+ α_S errors from the three groups.



Top studies within the NNPDF framework



Shortcomings of the Standard approach

What is the meaning of a one- σ uncertainty?

 Standard Δχ² = 1 criterion is too restrictive to account for large discrepancies among experiments.

[Collins & Pumplin, 2001]





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• Introduce a **TOLERANCE** criterion, i.e. take the envelope of uncertainties of experiments to determine the $\Delta \chi^2$ to use for the global fit (CTEQ).





Shortcomings of the Standard approach

What is the meaning of a one- σ uncertainty?

• Standard $\Delta \chi^2 = 1$ criterion is too restrictive to account for large discrepancies among experiments.

 Introduce a TOLERANCE criterion, i.e. take the envelope of uncertainties of experiments to determine the $\Delta \chi^2$ to use for the global fit

• Make it **DYNAMICAL**, i.e. determine $\Delta \chi^2$ separately for each hessian eigenvector

Y







(CTEQ).

(MSTW).

[Collins & Pumplin, 2001]

Shortcomings of the standard approach

What determines PDF uncertainties?

- Uncertainties in standard fits often increase when adding new data to the fit.
- Need of extending the parametrization in order to accomodate the new data

Smaller high-x gluon (and slightly smaller α_S) results in larger small-x gluon – now shown at NNLO.



Larger small-x uncertainty due to extrat free parameter.



PDF4LHCMSTW

A. Guffanti (Univ. Freiburg)

NNPDF Methodology

Main Ingredients

[R. D. Ball, L. Del Debbio, S. Forte, J. I. Latorre, J. Rojo, M. Ubiali and AG]

Monte Carlo determination of errors

- No need to rely on linear propagation of errors
- Possibility to test for non gaussianity of data
- Possibility to test for non-gaussian behaviour in fitted PDFs $(1 \sigma \text{ vs. } 68\% \text{ CL})$

Neural Networks

• Provide an unbiased parametrization

• Stopping based on Cross Validation

Ensures proper fitting avoiding overlearning



Dataset



3415 data points

(for comparison MSTW08 includes 2699 data points)

OBS	Data set			
Deep Inelastic Scattering				
F_2^d/F_2^p	F ^d ₂ /F ^p ₂ NMC-pd			
F_2^p	NMC			
_	SLAC			
	BCDMS			
F_2^d	SLAC			
_	BCDMS			
σ_{NC}^{\pm}	HERA-I comb.			
	ZEUS (HERA-II)			
σ_{CC}^{\pm}	HERA-I comb.			
	ZEUS (HERA-II)			
FL	H1			
$\sigma_{\nu}, \sigma_{\bar{\nu}}$	CHORUS			
dimuon prod.	NuTeV			
Drell-Yan & Vector Boson prod.				
$d\sigma^{\rm DY}/dM^2 dy$	E605			
$d\sigma^{\rm DY}/dM^2 dx_F$	E866			
W asymm.	CDF			
Z rap. distr.	D0/CDF			
Inclusive jet prod.				
Incl. $\sigma^{(jet)}$	CDF (k_T) - Run II			

Incl. $\sigma^{(jet)}$

D0 (cone) - Run II

S

- Fast DGLAP evolution based on higher-order interpolating polynomials
- Improved treatment of normalization errors (t₀ method)
 - For details see [R. D. Ball et al., arXiv:0912.2276]
- Improvements in training/stopping
 - Target Weighted Training
 - Improved stopping for avoiding under-/over-learning
- All details given in [R. D. Ball et al., arXiv:1002.4407]



FastKernel

- NLO computation of hadronic observables too slow for parton global fits.
- MSTW08 and CTEQ include Drell-Yan NLO as (local) K factors rescaling the LO cross section
- K-factor depends on PDFs and it is not always a good approximation.

- NNPDF2.0 includes full NLO calculation of hadronic observables.
- Use available fastNLO interface for jet inclusive cross-sections.[hep-ph/0609285]
- * Built up our own **FastKernel** computation of DY observables.

$$\int_{x_{0,1}}^{1} dx_1 \int_{x_{0,2}}^{1} dx_2 f_a(x_1) f_b(x_2) \mathcal{C}^{ab}(x_1, x_2) \rightarrow \sum_{\alpha, \beta=1}^{N_x} f_a(x_{1,\alpha}) f_b(x_{2,\beta}) \int_{x_{0,1}}^{1} dx_1 \int_{x_{0,2}}^{1} dx_2 \mathcal{I}^{(\alpha, \beta)}(x_1, x_2) \mathcal{C}^{ab}(x_1, x_2) = \sum_{\alpha, \beta=1}^{N_x} f_a(x_1, \alpha) f_b(x_2, \beta) \int_{x_{0,1}}^{1} dx_1 \int_{x_{0,2}}^{1} dx_2 \mathcal{I}^{(\alpha, \beta)}(x_1, x_2) \mathcal{C}^{ab}(x_1, x_2) = \sum_{\alpha, \beta=1}^{N_x} f_a(x_1, \alpha) f_b(x_2, \beta) \int_{x_{0,1}}^{1} dx_1 \int_{x_{0,2}}^{1} dx_2 \mathcal{I}^{(\alpha, \beta)}(x_1, x_2) \mathcal{C}^{ab}(x_1, x_2) = \sum_{\alpha, \beta=1}^{N_x} f_a(x_1, \alpha) f_b(x_2, \beta) \int_{x_{0,1}}^{1} dx_1 \int_{x_{0,2}}^{1} dx_2 \mathcal{I}^{(\alpha, \beta)}(x_1, x_2) \mathcal{C}^{ab}(x_1, x_2) = \sum_{\alpha, \beta=1}^{N_x} f_a(x_1, \beta) f_b(x_2, \beta) \int_{x_{0,1}}^{1} dx_1 \int_{x_{0,2}}^{1} dx_2 \mathcal{I}^{(\alpha, \beta)}(x_1, x_2) \mathcal{C}^{ab}(x_1, x_2) = \sum_{\alpha, \beta=1}^{N_x} f_a(x_1, \beta) f_b(x_2, \beta) \int_{x_{0,1}}^{1} dx_1 \int_{x_{0,2}}^{1} dx_2 \mathcal{I}^{(\alpha, \beta)}(x_1, x_2) \mathcal{C}^{ab}(x_1, x_2) = \sum_{\alpha, \beta=1}^{N_x} f_a(x_1, \beta) f_b(x_2, \beta) \int_{x_{0,2}}^{1} dx_1 \mathcal{L}^{(\alpha, \beta)}(x_1, x_2) \mathcal{L}^{(\alpha, \beta)}(x_1, x_2) \mathcal{L}^{(\alpha, \beta)}(x_1, x_2) = \sum_{\alpha, \beta=1}^{N_x} f_a(x_1, \beta) f_b(x_2, \beta) \int_{x_{0,2}}^{1} dx_1 \mathcal{L}^{(\alpha, \beta)}(x_1, x_2) \mathcal{L}$$

PDF and tops



- Both PDFs evolution and double convolution sped up by
 - Use high-orders polynomial interpolation
 - Precompute all Green Functions

A truly NLO analysis



General features of the fit





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PDF and tops

Partons - Comparison to other global fits



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PDF and tops

Results - Partons - A couple of upshots

 Reduction of uncertainties with respect to older NNPDF sets due to inclusion of new data





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• Uncertainties on PDFs competitive with results from other groups ...





Results - Partons - A couple of upshots

 Reduction of uncertainties with respect to older NNPDF sets due to inclusion of new data

• Uncertainties on PDFs competitive with results from other groups ...

 ... but still retain unbiasedness in regions where there are little or no experimental constraints



PDF Uncertainties and Correlations

A practitioner's guide to NNPDF predictions

Central Value

$$\langle \mathcal{F}
angle = rac{1}{N_{\text{set}}} \sum_{k=1}^{N_{\text{set}}} \mathcal{F}[q^{(k)}]$$



$\begin{aligned} \rho &\equiv \cos \varphi(\mathcal{F}, \mathcal{G}) = \frac{\langle \mathcal{F} \mathcal{G} \rangle_{\text{rep}} - \langle \mathcal{F} \rangle_{\text{rep}} \langle \mathcal{G} \rangle_{\text{rep}}}{\sqrt{\langle \mathcal{F}^2 \rangle_{\text{rep}} - \langle \mathcal{F} \rangle_{\text{rep}}^2} \sqrt{\langle \mathcal{G}^2 \rangle_{\text{rep}} - \langle \mathcal{G} \rangle_{\text{rep}}^2}} \end{aligned}$



$PDF - \sigma_{t\bar{t}}$ correlation



PDF-ttbar cross-section correlation

Mostly correlated with gluon distribution at medium-x



 $\sigma_{W^{\pm}} - \sigma_{t\bar{t}}$ correlation

	$\sigma_{\mathbf{W}^+}$	$\sigma_{\mathbf{W}^{-}}$	σz₀
ρ	-0.716	-0.694	-0.773



- Strong anti-correlation between the cross-sections
- Use the W^{\pm} , Z^0 cross-section (larger, known at NNLO) to normalize the $t\bar{t}$ cross-section

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 $PDF - \sigma_{t+X}$ (t-channel) correlation



PDF-single-top cross-section correlation

 Mostly correlated with gluon and quark distributions at medium-/small-x



 $\sigma_{W^{\pm}} - \sigma_{t+X}$ (t-channel)correlation

	$\sigma_{\mathbf{W}^+}$	$\sigma_{\mathbf{W}^{-}}$	σ z ٥
ρ	0.330	0.140	0.240



- Mild correlation between the cross-sections
- More difficult to use the Vector Boson cross-section as normalization



Assessing the impact of new data on PDF fits

- The N_{rep} replicas of a NNPDF fit give the probability density in the space of PDFs
- Expectation values for observables are Monte Carlo integrals

$$\langle \mathcal{F}[f_i(x, Q^2)]
angle = rac{1}{N_{rep}} \sum_{k=1}^{N_{rep}} \mathcal{F}\Big(f_i^{(net)(k)}(x, Q^2)\Big)$$

... the same is true for errors, correlations, etc.

 We can assess the impact of including new data in the fit updating the probability density distribution.

Assessing the impact of new data on PDF fits

[W. Giele and S. Keller, hep-ph/9803393]

According to Bayes Theorem we have

$$P_{\text{new}}(\lambda) = P(\lambda|x^e) = \frac{P(x^e|\lambda)P_{\text{init}}(\lambda)}{P(x^e)}, \quad P(x^e|\lambda) = e^{-\frac{\chi^2_{\text{new}}(\lambda)}{2}}$$

Monte Carlo integrals are now weighted sums

$$\langle \mathcal{F}[f_i(x, Q^2)] \rangle = \sum_{k=1}^{N_{rep}} w_k \mathcal{F}\left(f_i^{(net)(k)}(x, Q^2)\right)$$

where the weights are

$$\mathbf{W}_{k} = \frac{\mathbf{e}^{-\frac{1}{2}\chi^{2}_{\text{new}}(\lambda^{k})}}{\sum_{i=1}^{N_{\text{rep}}} \mathbf{e}^{-\frac{1}{2}\chi^{2}_{\text{new}}(\lambda^{i})}}$$



Proof-of-concept: Inclusive Jet data, reweighting vs. refitting

- Use DIS+DY-fit as prior probability distribution
- Add Tevatron Inclusive Jet data through refitting and through reweighting



Reweighting and refitting yield statistically equivalent results



Reweighting real data: W lepton asymmetry

- In the NNPDF2.0 fit we only included CDF W asymmetry data
- We evaluated W electron asymmetry with NNPDF20 1000 replicas set using DYNNLO

[Catani et al., arXiv:0903.2120].

- .. and included D0 W electron asymmetry data points through reweighting.
- Main impact on reduction of middle-*x* Valence uncertainty.
- No need of refitting



Conclusions and Outlook

- A precise determination of PDF with reliable error estimation is crucial to fully exploit the physics potential of LHC experiments
- Interesting overlap between PDF and top physics
 - Use PDF induced correlations to improve measurements
 - Impact of higher-order corrections
 - Use single-top to constrain the *b*-quark distribution ...

• NNPDF2.0 is the first global NNPDF fit

- Exact inclusion of NLO corrections
- No sign of strong tension among different datasets
- Next steps: improved treatment of Heavy Flavour contributions, inclusion of higher order contributions (NNLO-QCD, EW, ..)

