

News from NNPDF

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On behalf of the NNPDF Collaboration
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Standard Model Benchmarks at High-Energy Hadron Colliders
DESY (Zeuthen), June 15-17, 2011

NNPDF Methodology

Main Ingredients

- **Monte Carlo** determination of errors
 - No need to rely on linear propagation of errors
 - Possibility to test for the impact of non gaussianly distributed errors
 - Possibility to test for non-gaussian behaviour in fitted PDFs
($1 - \sigma$ vs. 68% CL)
- **Neural Networks**
 - Provide an **unbiased** parametrization
- **Stopping based on Cross-Validation**
 - Ensures proper fitting avoiding overlearning



NNPDF Methodology

... in a Nutshell

- Generate N_{rep} **Monte-Carlo replicas** of the experimental data
(sampling of the probability density in the space of data)
- Fit a set of Parton Distribution Functions on each replica
(sampling of the probability density in the space of PDFs)
- **Expectation values** for observables are **Monte Carlo integrals**

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

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



Reweighting PDFs

Assessing the impact of new data on PDF fits

[R. D. Ball et al., arXiv:1012.0836]

- Inspired by Giele and Keller [[hep-ph/9803393](#)]
- 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**

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

(... 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.



Reweighting PDFs

Assessing the impact of new data on PDF fits

- According to **Bayes Theorem** we have

$$\mathcal{P}_{\text{new}}(\{f\}) = \mathcal{N}_x \mathcal{P}(\chi^2 | \{f\}) P_{\text{init}}(\{f\}), \quad \mathcal{P}(\chi^2 | \{f\}) = [\chi^2(y, \{f\})]^{\frac{n_{\text{dat}} - 1}{2}} e^{-\frac{\chi^2(y, \{f\})}{2}}$$

- Monte Carlo integrals** are now **weighted sums**

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

where the **weights** are

$$w_k = \frac{[\chi^2(y, f_k)]^{\frac{n_{\text{dat}} - 1}{2}} e^{-\frac{\chi^2(y, f_k)}{2}}}{\sum_{i=1}^{N_{\text{rep}}} [\chi^2(y, f_i)]^{\frac{n_{\text{dat}} - 1}{2}} e^{-\frac{\chi^2(y, f_i)}{2}}}$$



NNPDF2.2

Including Tevatron & LHC W asymmetry data in
NNPDF fits



NNPDF2.1@NNLO



NNPDF2.1@NNLO

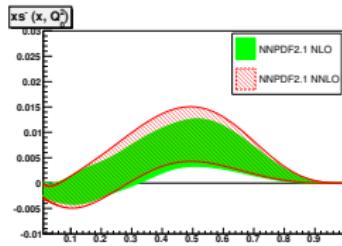
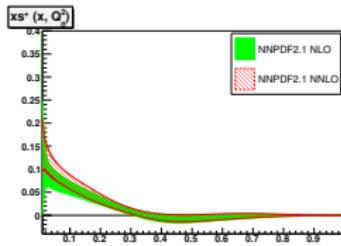
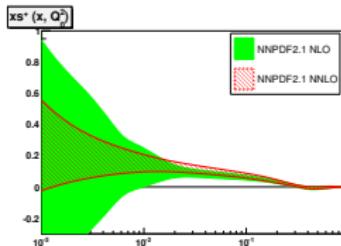
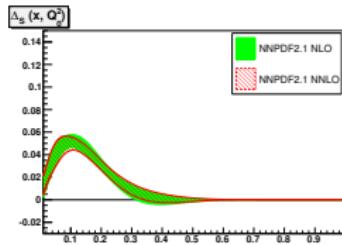
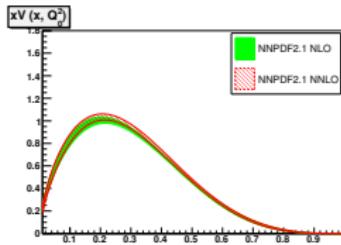
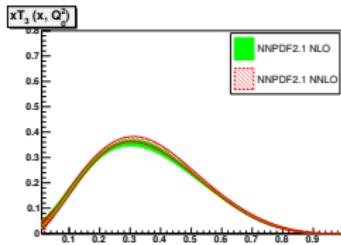
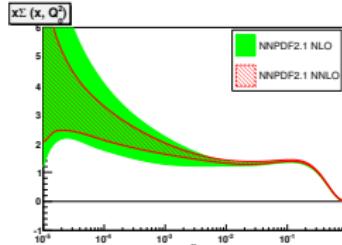
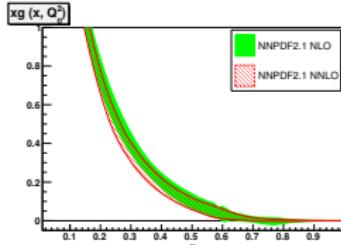
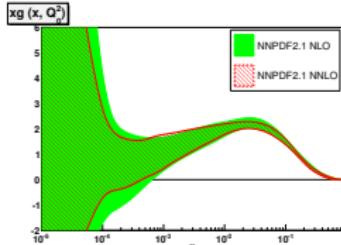
General Setup

- Same **dataset** as **NNPDF2.1 NLO**
(DIS, Drell-Yan, EWB and Inclusive Jet production at colliders).
- **Heavy quark mass effects** treated in the **FONLL-C** General Mass Variable Falavour Number Scheme.
($\mathcal{O}(\alpha_s^2)$ massless + $\mathcal{O}(\alpha_s^2)$ massive contributions).
- **NNLO corrections** to fixed-target **Drell-Yan** and Electroweak Vector Boson production cross-sections implemented as **local NNLO/NLO K-factors**.
- PDFs at the initial scale are parametrized with the **same redundant paramertrization** used in the NLO fits (**259 parameters**)



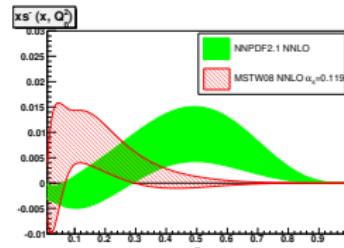
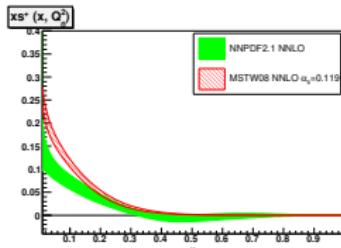
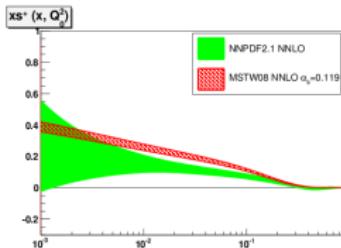
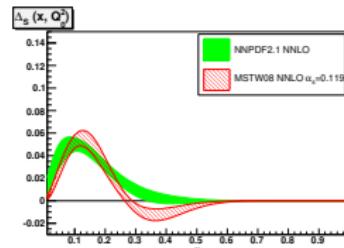
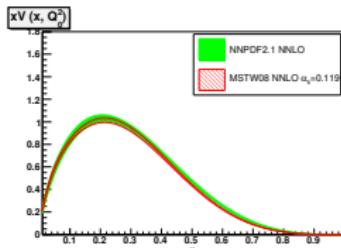
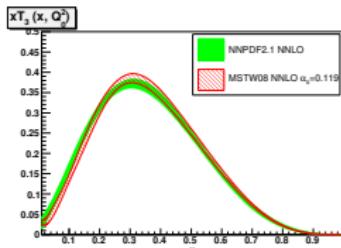
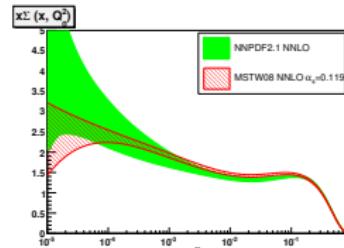
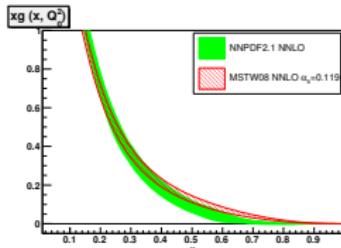
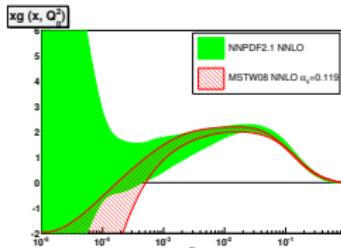
NNPDF2.1@NNLO

Partons - Comparison to NLO



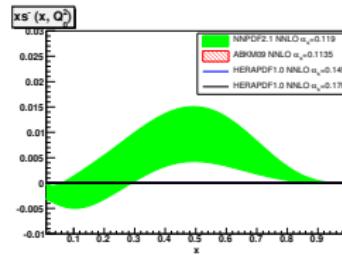
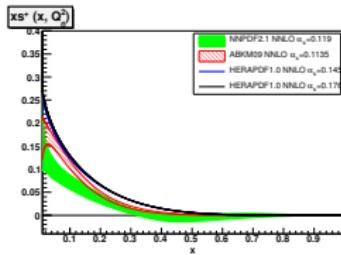
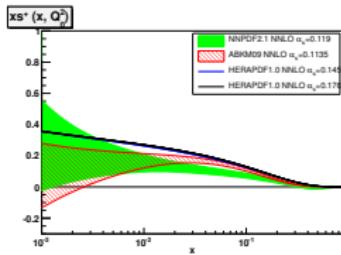
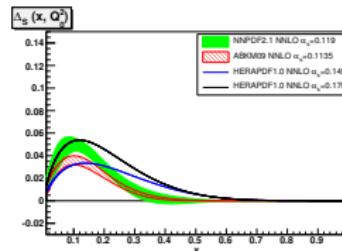
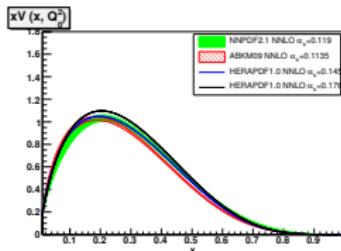
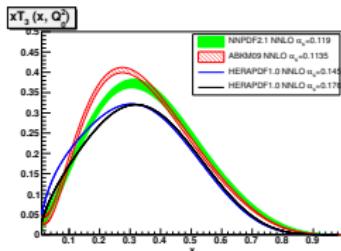
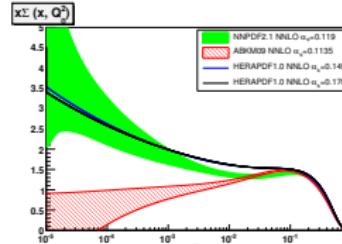
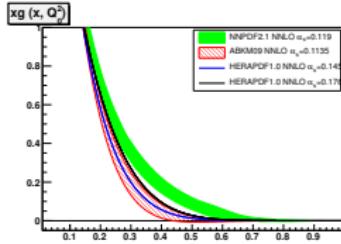
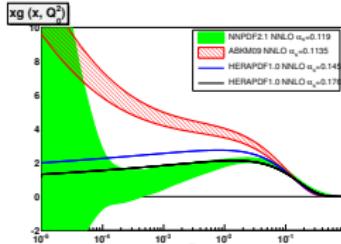
NNPDF2.1@NNLO

Partons - Comparison to MSTW08



NNPDF2.1@NNLO

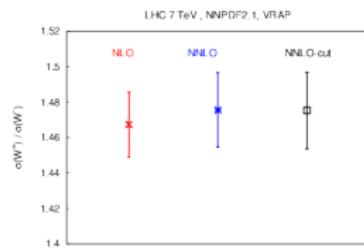
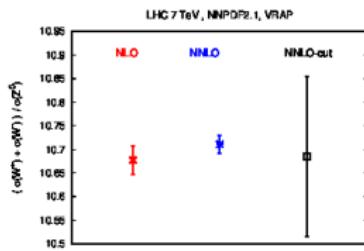
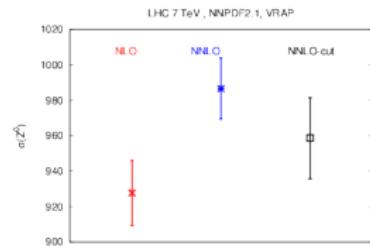
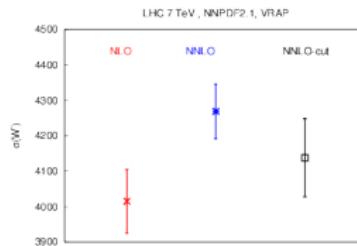
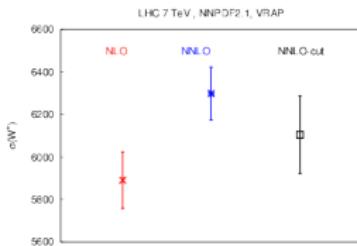
Partons - Comparison to ABKM09 and HERAPDF1.0



NNPDF2.1@NNLO

Phenomenology - LHC Standard Candles

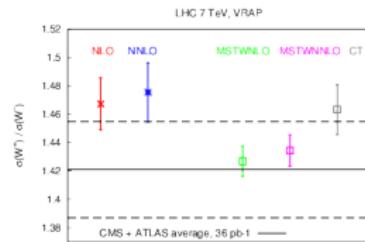
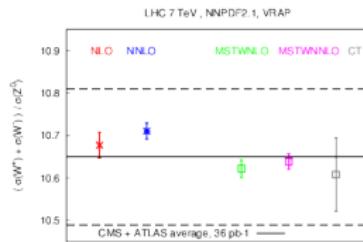
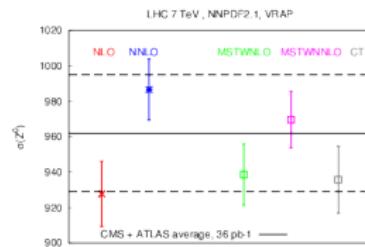
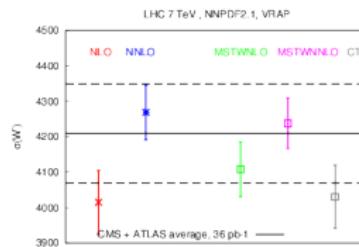
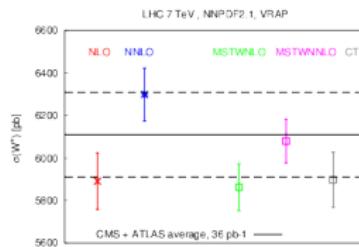
- Predictions for **LHC Standard Candles** compared to **NLO results**



NNPDF2.1@NNLO

Phenomenology - LHC Standard Candles

- Predictions for **LHC Standard Candles** compared to **LHC data**



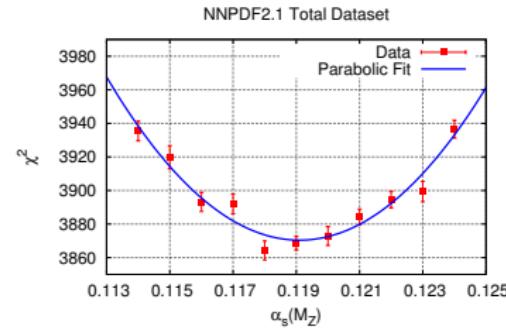
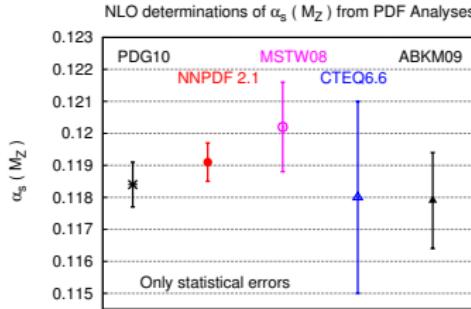
$\alpha_s(M_Z)$ determination from NNPDF analysis



$\alpha_s(M_Z)$ from PDF analysis

[S. Lionetti et al., arXiv:1103.2369]

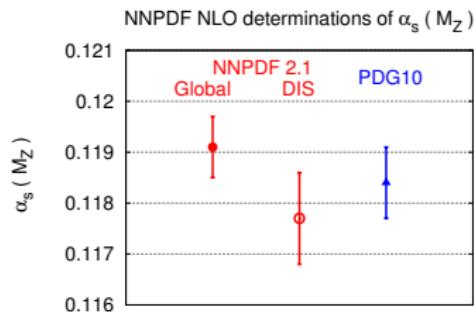
- **The Good:** Large dataset
⇒ Small statistical errors
- **The Bad:** Best fit α_S and PDFs are correlated
⇒ Parametrization bias? Dataset Dependence?
- **The Ugly:** Need to tame statistical fluctuations in χ^2
⇒ Large replica samples!



$\alpha_s(M_Z)$ from PDF analysis

Dependence of $\alpha_s(M_Z)$ on the dataset

	$\alpha_s(M_Z)$
NNPDF2.1	$0.1191 \pm 0.0006^{\text{stat}}$
NNPDF2.1 DIS-only	$0.1177 \pm 0.0009^{\text{stat}}$
NNPDF2.0	$0.1168 \pm 0.0007^{\text{stat}}$
NNPDF2.0 DIS-only	$0.1145 \pm 0.0010^{\text{stat}}$

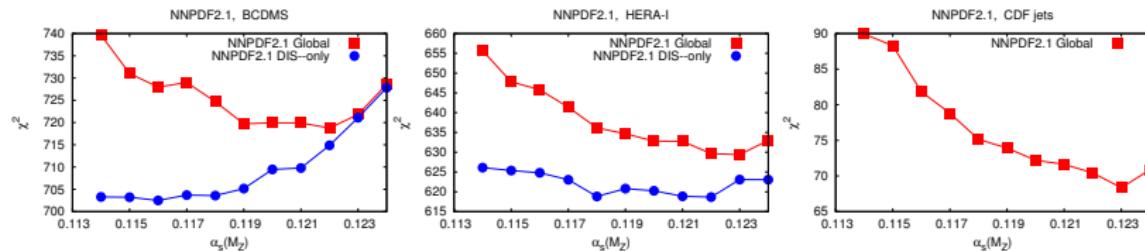


- Do DIS data prefer a smaller value of α_s ?
Maybe (but not much smaller), and anyway **compatible with the value from global fit and with larger uncertainties.**
- Theoretical uncertainties likely dominant (Ex. $\Delta\alpha_s^{\text{HQ}} \sim 0.002$)



$\alpha_s(M_Z)$ from PDF analysis

$\alpha_s(M_Z)$ from individual experiments



- BCDMS in a DIS-only fit sometimes has runaway direction at small $\alpha_s(M_Z)$, absent in the global fit
- HERA rather flat in α_s in DIS-only fit
- Tevatron jet experiments exclude small $\alpha_s(M_Z)$ values

Significant interplay between DIS and hadronic data



Conclusions

- Parton Distribution Functions are a fundamental ingredient for **precision LHC phenomenology**.
- The NNPDF Methodology combining Monte Carlo techniques and Neural Networks provides a way to address problems affecting standard fits.
- Inclusion of **Tevatron and LHC W asymmetry data in global fit** possible without generating a tension with other datasets \Rightarrow **NNPDF2.2** available soon in LHAPDF.
- **NNLO** is necessary for comparison to **high precision Benchmark measurements** \Rightarrow **NNPDF2.1@NNLO** analysis almost complete, available in LHAPDF soon.
- Extraction of strong coupling constant from NNPDF analysis performed at NLO. NNLO will come in due time.



BACKUP SLIDES



PDF Uncertainties and Correlations

A practitioner's guide to NNPDF predictions

Central Value

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

Standard Deviation

$$\sigma_{\mathcal{F}} = \left(\frac{1}{N_{\text{set}}} \sum_{k=1}^{N_{\text{set}}} \left(\mathcal{F}[q^{(k)}] - \langle \mathcal{F} \rangle \right)^2 \right)^{1/2}$$

Correlation

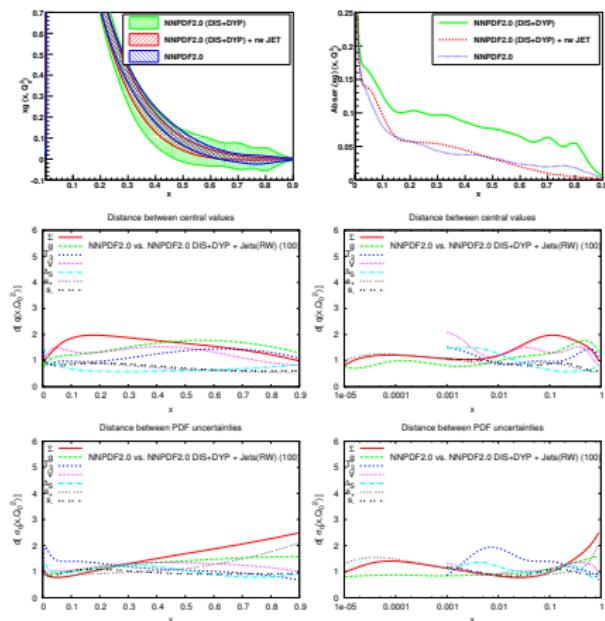
$$\rho \equiv \cos \varphi(\mathcal{F}, \mathcal{G}) = \frac{\langle \mathcal{F} \mathcal{G} \rangle_{\text{rep}} - \langle \mathcal{F} \rangle_{\text{rep}} \langle \mathcal{G} \rangle_{\text{rep}}}{\sqrt{\langle \mathcal{F}^2 \rangle_{\text{rep}} - \langle \mathcal{F} \rangle_{\text{rep}}^2} \sqrt{\langle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



Reweighting PDFs

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

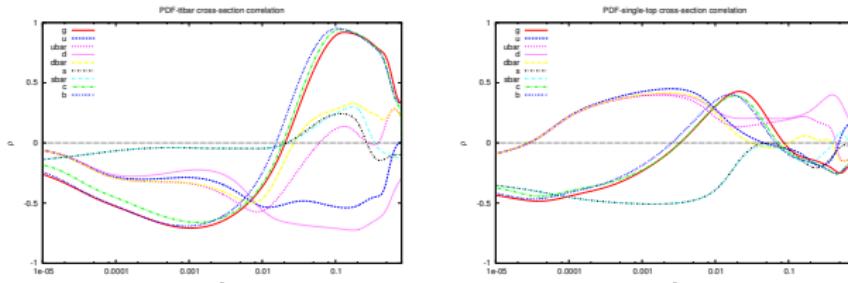


PDF induced correlations

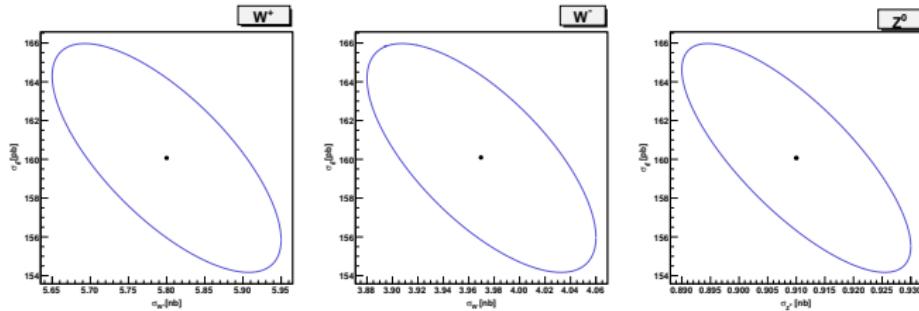
Ex.: Top-quark studies within the NNPDF framework

[J. Rojo and AG, arXiv:1008.4671]

- It is easy to compute **correlations** between **PDFs and observables**



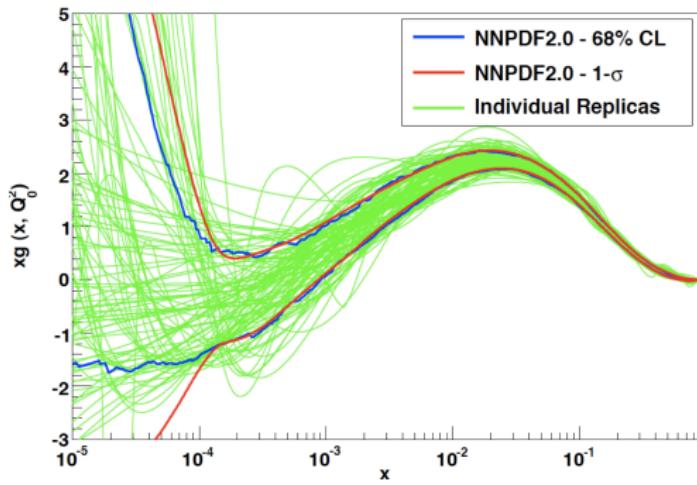
- ... or **pairs of observables**



Confidence Level Intervals

Testing for non gaussian distribution of fitted PDFs

- **Confidence Level intervals** can be computed directly from the replicas distribution
- Comparison of 68% C.L. and symmetric 1σ especially in extrapolation regions where theory constraints dominate on experimental information



NNPDF 2.1

FONLL - The gory details

- A generic DIS observable in the FONLL scheme is written as:

$$F^{FONLL}(x, Q^2) = \mathcal{D}(Q^2) F^{(d)}(x, Q^2) + F^{(n_l)}(x, Q^2)$$

where the **threshold damping factor** is given by

$$\mathcal{D}(Q^2) = \theta(Q^2 - m^2) \left(1 - \frac{m^2}{Q^2}\right)$$

and the **subtraction term** is

$$F^{(d)} = [F^{(n_l+1)}(x, Q^2) - F^{(n_l,0)}(x, Q^2)]$$

with the massless limit of the massive contributions being

$$F^{n_l,0}(x, Q^2) = x \int_x^1 \frac{dy}{y} \sum_{i=q,\bar{q},g} B_i^{(0)} \left(\frac{x}{y}, \frac{Q^2}{m^2}, \alpha_S^{(n_l+1)}(Q^2)\right) f_i^{(n_l+1)}(y, Q^2)$$

with

$$\lim_{m \rightarrow 0} \left[B_i \left(x, \frac{Q^2}{m^2}\right) - B_i^{(0)} \left(x, \frac{Q^2}{m^2}\right) \right] = 0$$

