

# Parton distributions with LHC data

## The new generation of PDFs

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The logo for NNPDF features the letters 'NNPDF' in a bold, blue, sans-serif font. The first two 'N's are stylized with a network of nodes and connections. The nodes are small white circles, and the connections are thin lines in blue and green, forming a complex web-like structure that suggests a neural network or a data-driven model.

in collaboration with S. Forte, R. Ball, L. Del Debbio *et al.*



- 1 Introduction
  - Definition
  - Motivation
- 2 NNPDF approach
  - Overview on current PDF providers
  - The NNPDF methodology
- 3 NNPDF with LHC data
  - The new dataset
  - PDF results with LHC data
  - Comparing results between different PDF providers
  - Phenomenology at  $\sqrt{s} = 8$  TeV
- 4 Conclusion and outlook



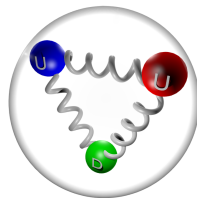
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- What are **parton distribution functions**?

## PDF definition:

The **probability density** of finding a constituent of the proton, called **parton**, with a momentum fraction  $x$  of the proton at momentum transfer  $Q^2$ .



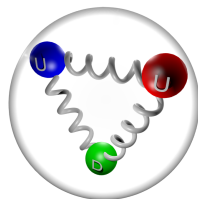
- **Partons** are **gluons**, **quarks** and **antiquarks**  
 $\Rightarrow g, u, \bar{u}, d, \bar{d}, s, \bar{s}, \dots$



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- **Partons** are **gluons**, **quarks** and **antiquarks**  
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- Facts about PDFs:
  - ▶ **Not calculable**: reflect **non-perturbative** physics of confinement
  - ▶ Extracted by comparing **theoretical predictions** to **experimental** data:

Deep Inelastic  
Scattering

Drell-Yan  
process

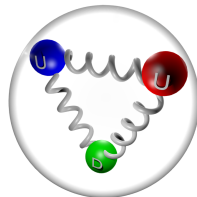
Electroweak  
distributions

Jet  
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PDF  $\neq$



# Motivation - Computing observables

- Why do we need **parton distribution functions**?

Following the factorization theorem in QCD, the observable of a hard process ( $Q \gg \Lambda_{\text{QCD}}$ ) can be written as

$$\frac{d\sigma}{dx dQ^2} = \sum_{i=1}^{n_f} \frac{d\hat{\sigma}_i}{dQ^2} \otimes f_{i/p}(x, Q^2),$$

the sum over all PDF flavors  $n_f$  of the convolution product between:

$$\frac{d\hat{\sigma}_i}{dQ^2}$$

the **elementary “hard” cross section** which is computed in QCD, depends on the physical process.

$\otimes$

$$f_{i/p}(x, Q^2)$$

the **PDF of parton  $i$**  inside a proton  $p$ , carrying a momentum fraction  $x$  at the energy scale  $Q^2$ .



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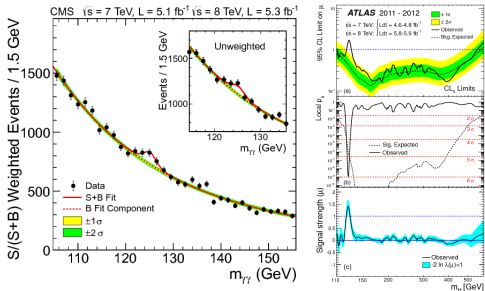
the **PDF of parton  $i$**  inside a proton  $p$ , carrying a momentum fraction  $x$  at the energy scale  $Q^2$ .

**Remark:** PDFs are essential for theoretical predictions.



# Motivation - Using PDFs

- Where do we need **parton distribution functions**?

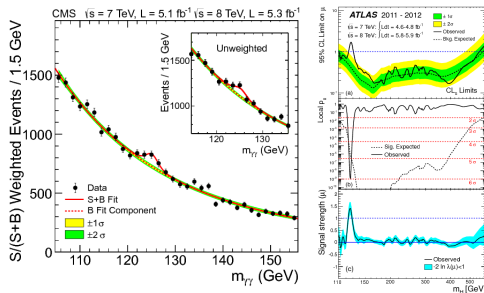


- PDFs are necessary to determine theoretical predictions for signal/background of experimental measurements.



# Motivation - Using PDFs

- Where do we need **parton distribution functions**?



- PDFs are necessary to determine theoretical predictions for signal/background of experimental measurements.

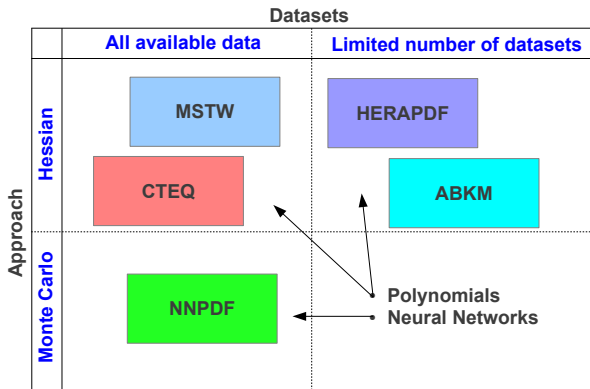
Problem: PDF **uncertainties** on signal range from **4% up to 8%**, accordingly to the channel, of the systematic uncertainties of the measurements.

Solution: Improve PDF extraction **methodology** and add more **data** (LHC data).

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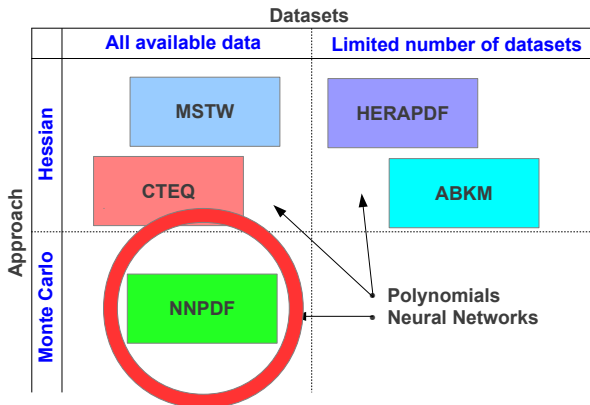
# PDFs on the market...



- The providers use different **approaches** and **datasets**.
- **NNPDF** uses a high technological approach.



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# The NNPDF methodology (shortly)

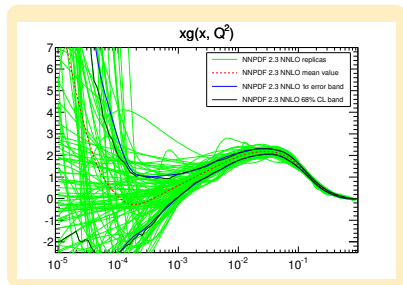
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  - ▶ implementation started in 2002, based on **Neural Networks**,
  - ▶ reduction of all sources of theoretical **bias**, e.g. the functional form.

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- PDFs are extracted by using
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  - ▶ Optimization controlled by **training/validation method**,
  - ▶ **Monte Carlo** representation of results.

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  - ▶ **Monte Carlo** representation of results.
- Expectation values for **observables** are **Monte Carlo integrals**:



MC approach:

$$\langle \mathcal{O}[f] \rangle = \frac{1}{N} \sum_{k=1}^N \mathcal{O}[f_k]$$

- ▶ for **observables**, **errors**, **correlations**, etc.



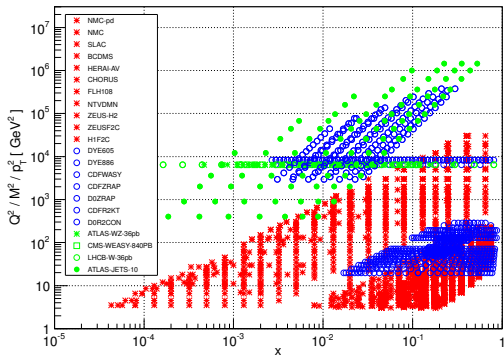
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# NNPDF2.3 - The new dataset

- **NNPDF2.3** is the first PDF set which includes **all the LHC data** for which the complete experimental information is available (●).

NNPDF2.3 dataset



## ATLAS:

- inclusive jets
- W/Z lepton rapidity distributions

## CMS:

- W lepton asymmetry

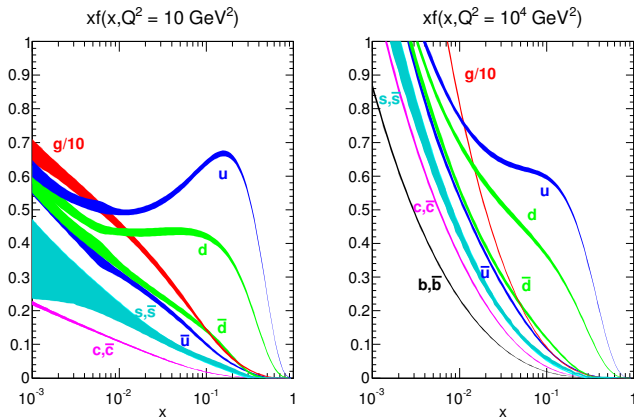
## LHCb:

- W rapidity distributions

- **3501 data points: 51 LHC W/Z, 90 LHC Jets**

(ATLAS jets arXiv:1112.6297, ATLAS W/Z arXiv:1109.5141, CMS Weasy arXiv:1206.2598, LHCb W arXiv:1204.1620)

## NNPDF2.3 NLO PDFs (1- $\sigma$ std. dev.)



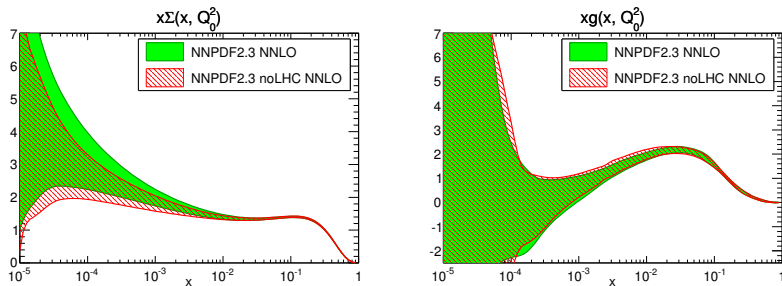
● NNPDF2.3 sets are also available:

- ▶ varying datasets,
- ▶ varying QCD parameters, e.g.  $\alpha_s$ ,
- ▶ with different theoretical frameworks, e.g. perturbative order (NLO/NNLO).



# Example of NNPDF2.3 PDF

- Example of PDFs at  $Q_0^2 = 2.0 \text{ GeV}^2$ , with and without LHC data.



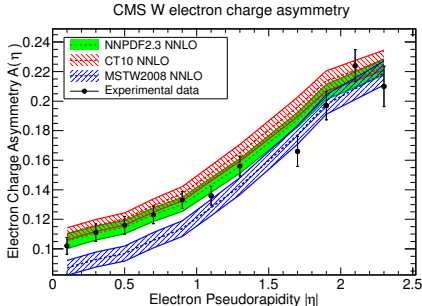
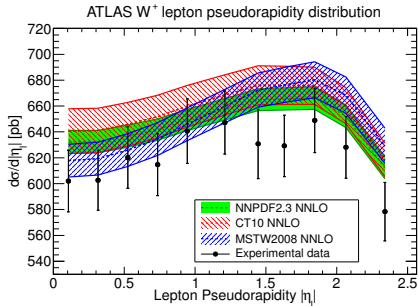
- LHC data reduces **PDF uncertainties** and **improves the  $\chi^2/\text{d.o.f}$**  between data and theoretical predictions produced by the PDFs:

	<b>NNPDF2.3 noLHC</b>	<b>NNPDF2.3</b>
Total $\chi^2/\text{d.o.f.}$	<b>1.142</b>	<b>1.139</b>

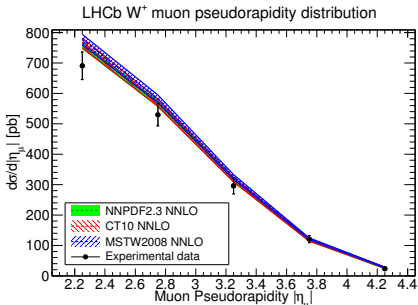
- Improvement about 10 units of  $\chi^2$ .



# Example of LHC observables

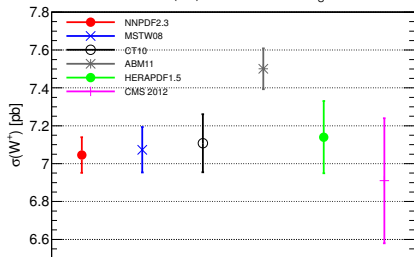


NNLO, $\alpha_S = 0.119$			
PDF Set	NNPDF2.3	MSTW08	CT10
ATLAS W, Z	1.435	3.201	1.160
CMS Weasy	0.813	3.862	1.772
LHCb W	0.831	1.050	0.966
ATLAS jets	0.937	0.935	1.016

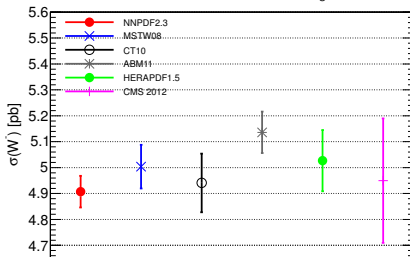


# Candle Cross Sections - $W^\pm$ and $Z$

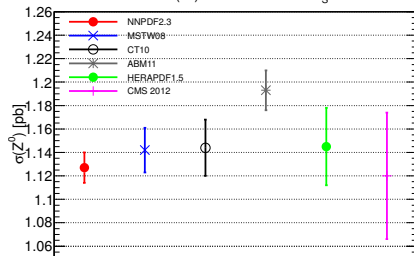
LHC 8 TeV  $\sigma(W^\pm)$  - VRAP NNLO -  $\alpha_s = 0.119$



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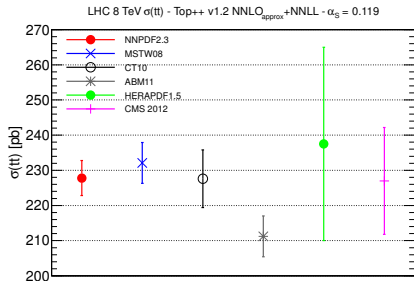
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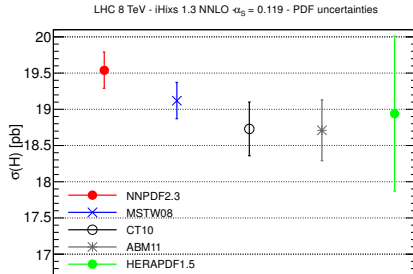
Sets extracted from all datasets produce similar and more precise results (**NNPDF/CT10/MSTW**)

Sets extracted from partial datasets have large uncertainties and less precision (**ABM/HERAPDF**).

# Candle Cross Sections - Higgs and Top



(a) Top inclusive cross section.



(b) Higgs inclusive cross section.

- Waiting for **Higgs** cross section measurements!



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# Conclusion and outlook

- 1 NNPDF2.3 is the **first PDF set** including **all the LHC data**
- 2 The impact of LHC data is small but **non-negligible**
- 3 The paper will be impressed by Nuclear Physics B.

## Outlook:

- Improvements of the code and the general structure.
- Near future: Electroweak corrections to PDFs, **photon** PDF.



## BACKUP SLIDES

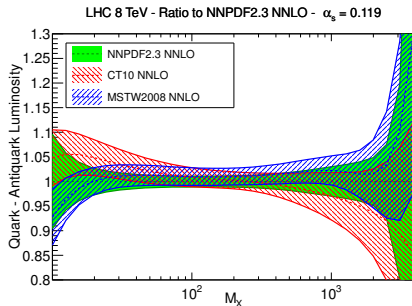
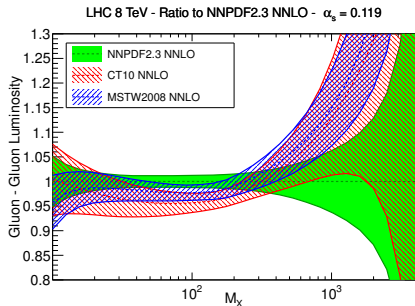


# Luminosities at $\sqrt{s} = 8$ TeV

**Luminosity** is defined as

$$\Phi_{ij}(M_X^2) = \frac{1}{s} \int_{\tau}^1 \frac{dx_1}{x_1} f_i(x_1, M_X^2) f_j(\tau/x_1, M_X^2), \quad \tau = \frac{M_X^2}{s}$$

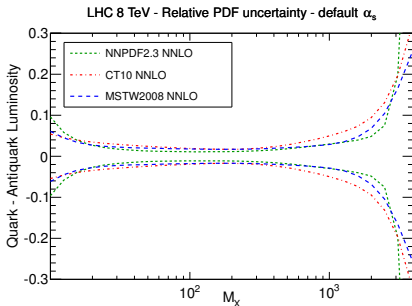
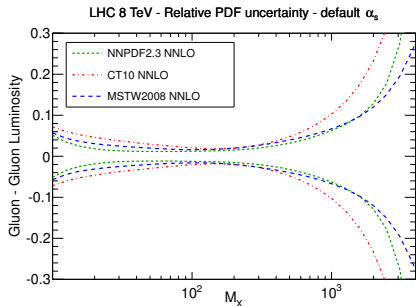
- Preliminary results:  $gg$  and  $q\bar{q}$  luminosities at **8 TeV** (2012 runs)



- All luminosities are reasonably compatible between global PDF sets.



- Results:  $gg$  and  $q\bar{q}$  luminosities PDF relative uncertainty at **8 TeV**.



- NNPDF2.3 with LHC data has **small relative uncertainties**  
⇒ Improvement of **quality** in **theoretical predictions**.



- Heavy quark mass effects included using the FONLL method up to NNLO, S. Forte et al., arXiv:1001.2312.
- FastKernel method for the inclusion of the higher order corrections  
The NNPDF Collaboration, arXiv:1002.4407
  - ▶ DIS up to NNLO
  - ▶ DY and JET up to NLO
- NNLO corrections to DY included by means of K-factors (DYNNLO)
- NNLO corrections to inclusive JET implement using FastNLO (hep-ph/0609285)
  - ▶ approximated NNLO corrections based on threshold resummation.



- NLO/NNLO cuts

- ▶  $W^2 = Q^2(1-x)/x > 12.5 \text{ GeV}^2$
- ▶  $Q^2 > 3 \text{ GeV}^2$  + further cuts on  $F_2^c$

- ATLAS W, Z lepton rapidity distributions cuts

$$p_T^l \geq 20 \text{ GeV}, \quad p_T^{\nu} \geq 25 \text{ GeV}, \quad m_T < 40 \text{ GeV}, \quad |\eta_l| \leq 2.5$$

$$p_T^l \geq 20 \text{ GeV}, \quad 66 \text{ GeV} \leq m_{l+l-} \leq 116 \text{ GeV}, \quad \eta_{l+,l-} \leq 4.9$$

- CMS W electron asymmetry

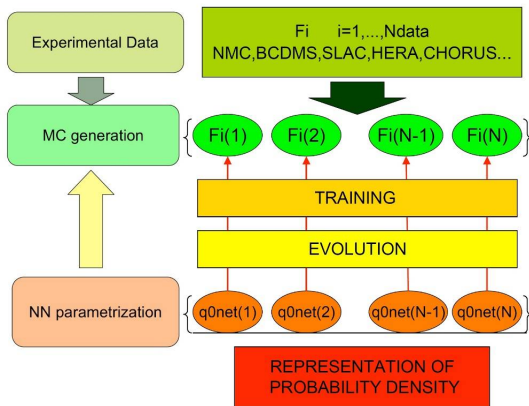
$$p_T^e \geq 35 \text{ GeV}$$

- LHCb W, Z rapidity distribution

$$p_T^{\mu} \geq 20 \text{ GeV}, \quad 60 \text{ GeV} \leq m_{l+l-} \leq 120 \text{ GeV}, \quad 2.0 \leq \eta_{1,2}^{\mu} \leq 4.5$$



# NNPDF mechanism



# Neural Networks

- Neural Networks are defined as

$$\xi_i^{(l)} = g \left( \sum_{j=1}^{n_{l-1}} \omega_{ij}^{(l-1)} \xi_j^{(l-1)} - \theta_i^{(l)} \right)$$

- $\omega_{ij}^{(l-1)}$  weights,  $\theta_i^{(l)}$  thresholds,  $i^{th}$  neuron,  $l^{th}$  layer, where  $g$  is the sigmoid activation function:

$$g(x) \equiv \frac{1}{1 + e^{-x}}$$

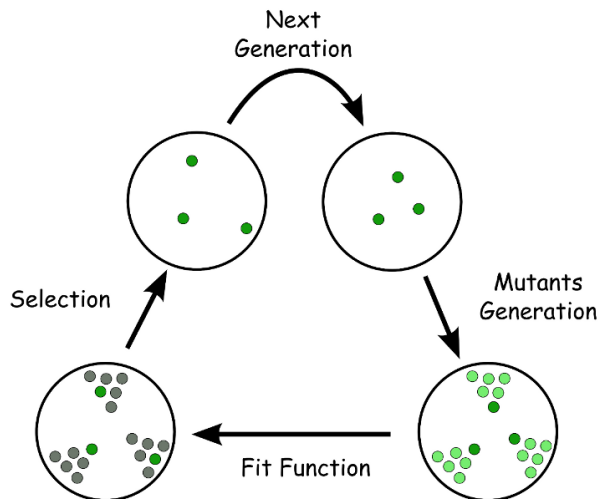
- Example: Neural network 1-2-1

$$\xi_1^{(3)} = \left\{ 1 + \exp \left[ \theta_1^{(3)} - \frac{\omega_{11}^{(2)}}{1 + e^{\theta_1^{(2)} - x\omega_{11}^{(1)}}} - \frac{\omega_{12}^{(2)}}{1 + e^{\theta_2^{(2)} - x\omega_{21}^{(1)}}} \right] \right\}^{-1}$$





# Genetic algorithm



# Training/validation method

