

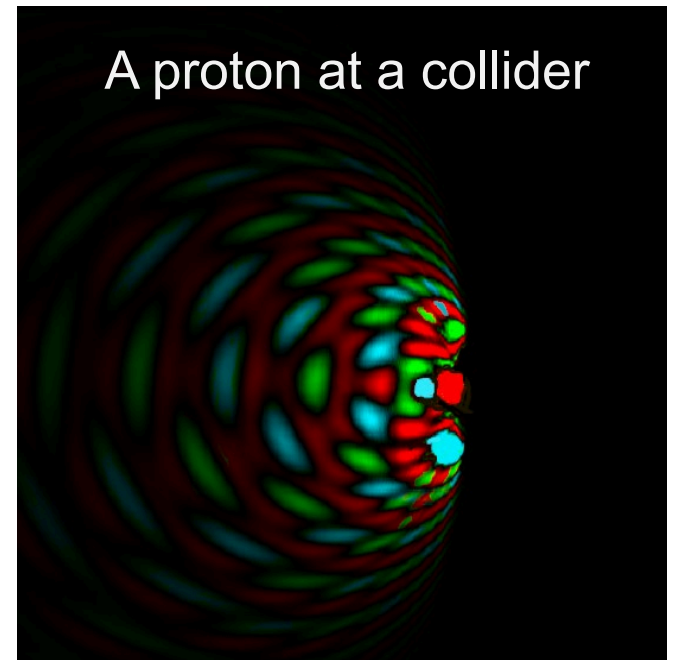
# Proton structure for precise QCD calculations

Balancing precision and replicability in uncertainty quantification

**Pavel Nadolsky**

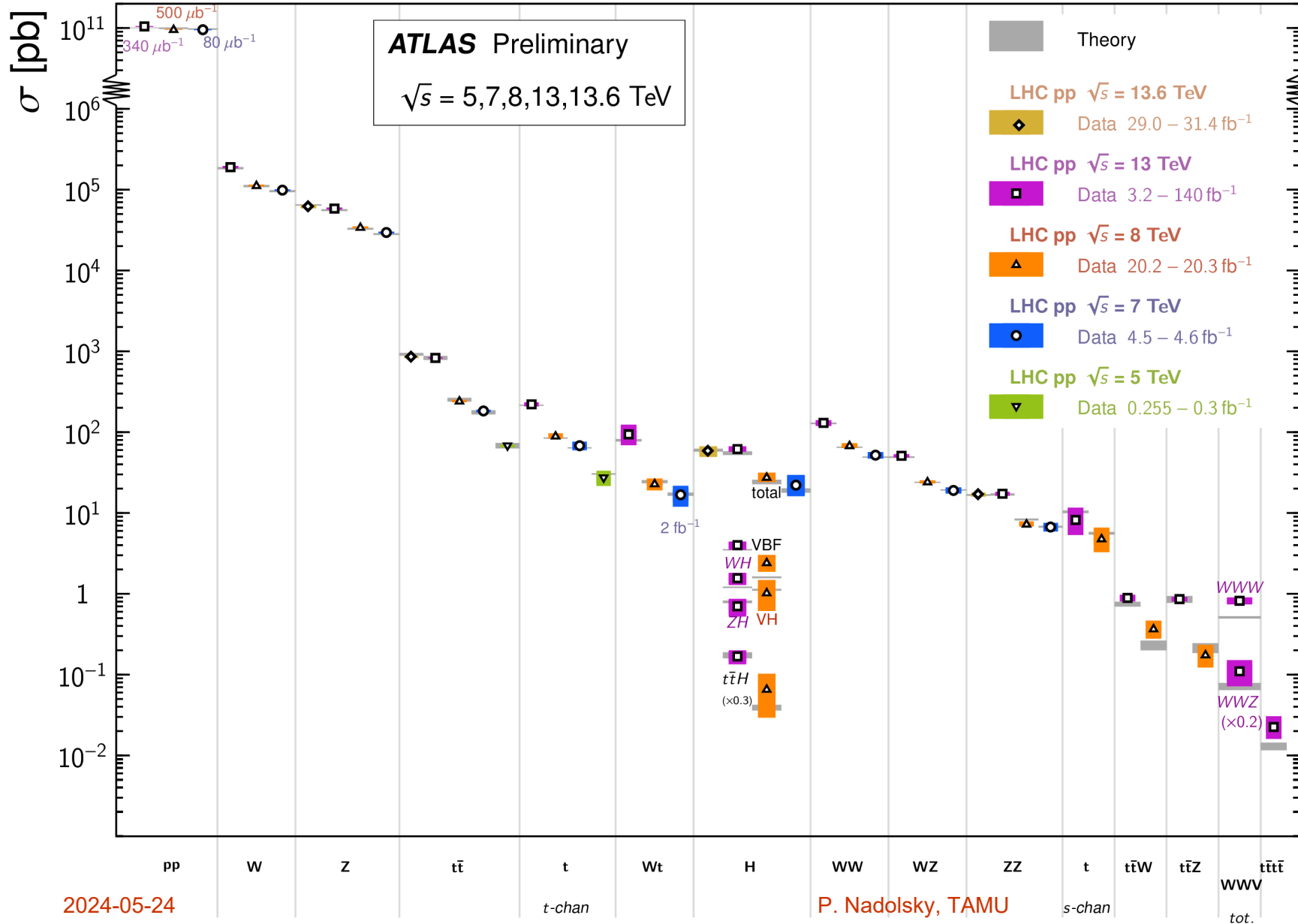
**Southern Methodist University**

With A. Courtoy, L. Kotz, F. Olness,  
and CTEQ-TEA (Tung Et Al.) Global QCD  
analysis group



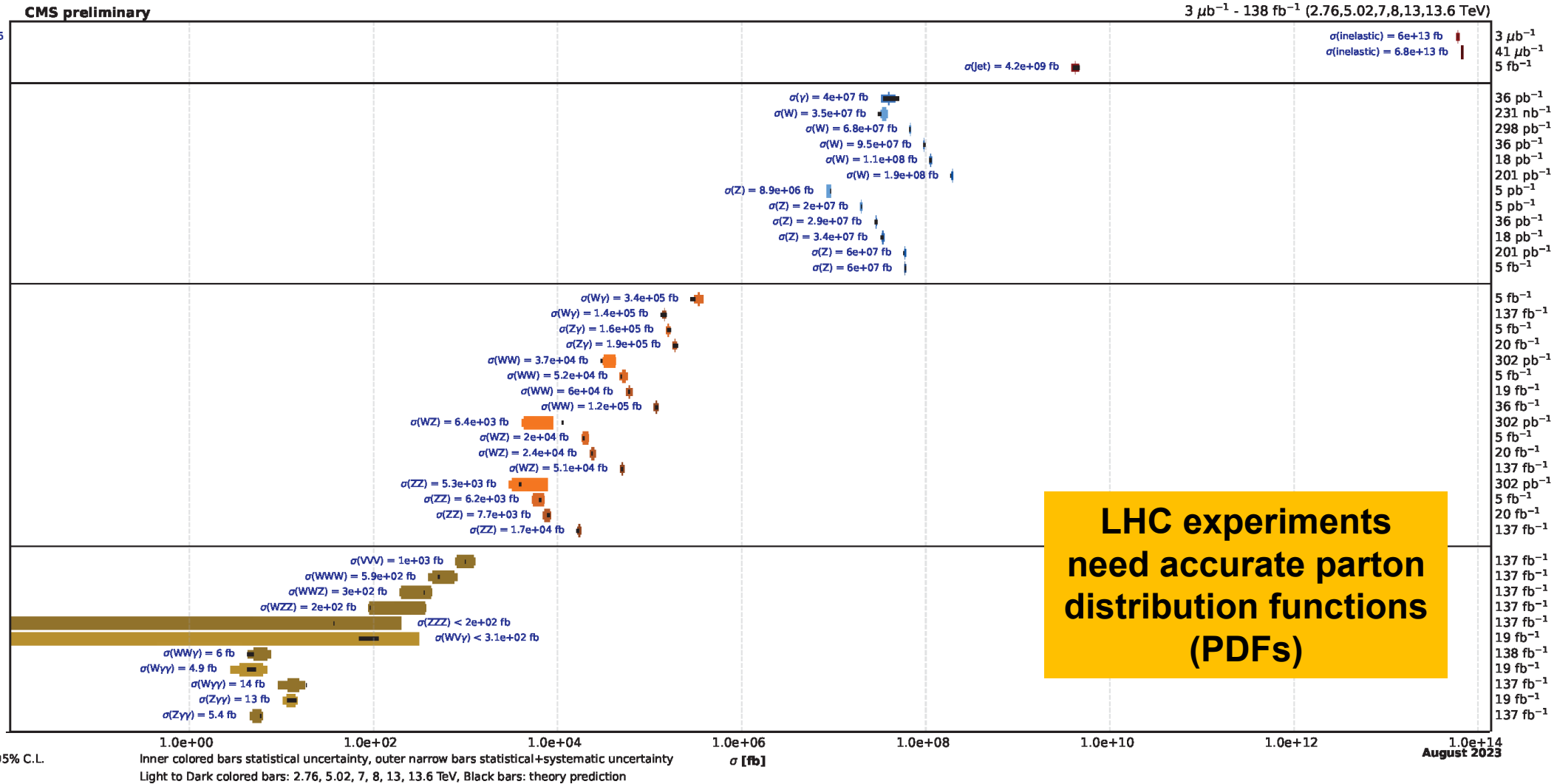
# Standard Model Total Production Cross Section Measurements

Status: October 2023



**LHC experiments need accurate parton distribution functions (PDFs)**

# Overview of CMS cross section measurements I



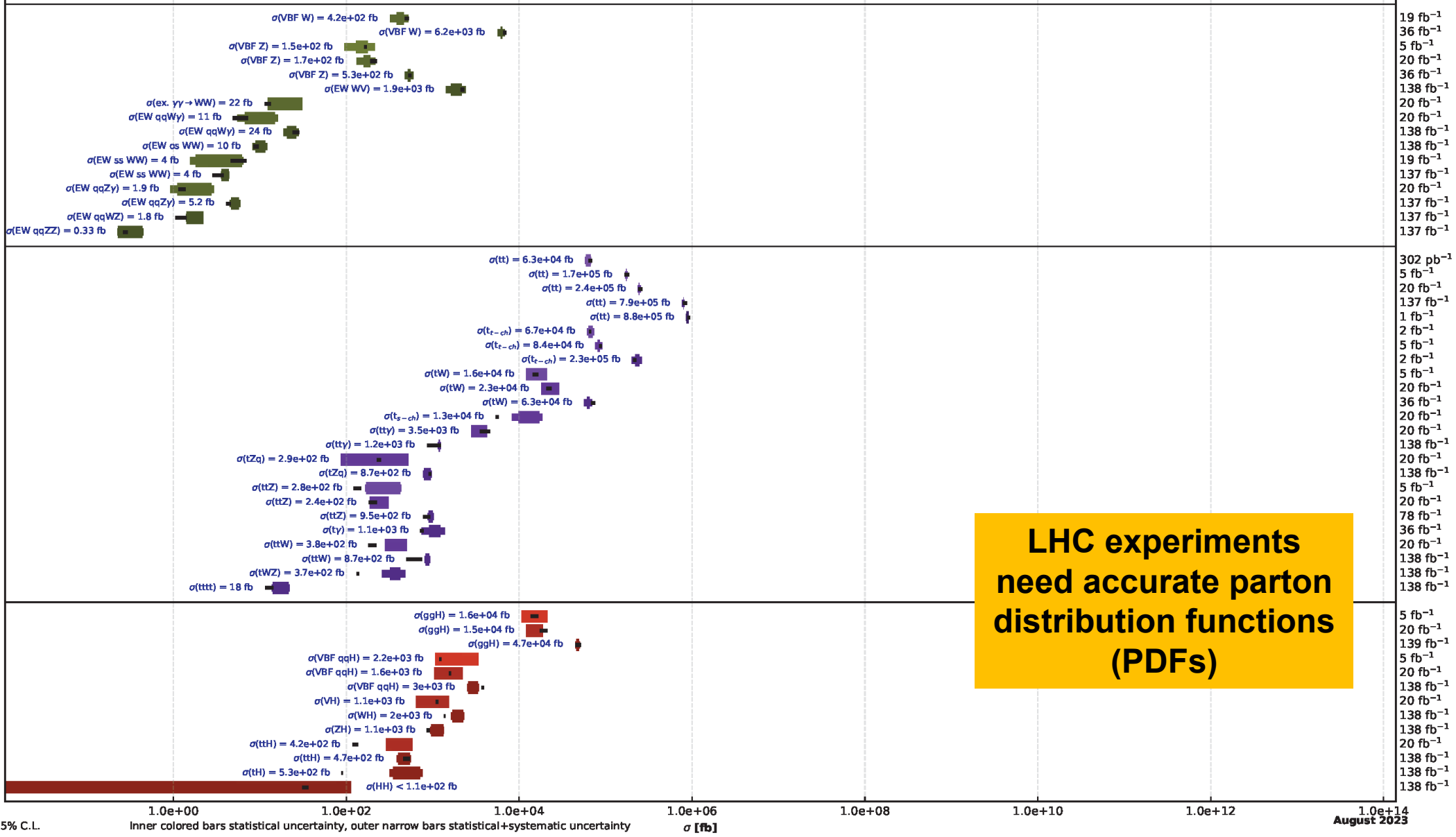
Measured cross sections and exclusion limits at 95% C.L.  
See here for all cross section summary plots

# Overview of CMS cross section measurements II

VBF W	8 TeV	JHEP 11 (2016) 147
VBF W	13 TeV	EPJC 80 (2020) 43
VBF Z	7 TeV	JHEP 10 (2013) 101
VBF Z	8 TeV	EPJC 75 (2015) 66
VBF Z	13 TeV	EPJC 78 (2018) 589
EW WW	13 TeV	PLB 834 (2022) 137438
ex. $\gamma\gamma \rightarrow WW$	8 TeV	JHEP 08 (2016) 119
EW qqW $\gamma$	8 TeV	JHEP 06 (2017) 106
EW qqW $\gamma$	13 TeV	PRD 108 032017
EW os WW	13 TeV	PLB 841 (2023) 137495
EW ss WW	8 TeV	PRL 114 051801 (2015)
EW ss WW	13 TeV	PLB 809 (2020) 135710
EW qqZ $\gamma$	8 TeV	PLB 770 (2017) 380
EW qqZ $\gamma$	13 TeV	PRD 104 072001 (2021)
EW qqWZ	13 TeV	PLB 809 (2020) 135710
EW qqZZ	13 TeV	PLB 812 (2020) 135992

tt	5.02 TeV	JHEP 04 (2022) 144
tt	7 TeV	JHEP 08 (2016) 029
tt	8 TeV	JHEP 08 (2016) 029
tt	13 TeV	PRD 104 (2021) 092013
tt	13.6 TeV	Submitted to JHEP
$t\bar{t}-ch$	7 TeV	JHEP 12 (2012) 035
$t\bar{t}-ch$	8 TeV	JHEP 06 (2014) 090
$t\bar{t}-ch$	13 TeV	PLB 72 (2017) 752
tW	7 TeV	PRL 110 (2013) 022003
tW	8 TeV	PRL 112 (2014) 231802
tW	13 TeV	JHEP 10 (2018) 117
$t\bar{t}-ch$	8 TeV	JHEP 09 (2016) 027
tty	8 TeV	JHEP 10 (2017) 006
tty	13 TeV	JHEP 05 (2022) 091
tZq	8 TeV	JHEP 07 (2017) 003
tZq	13 TeV	JHEP 02 (2022) 107
ttZ	7 TeV	PRL 110 (2013) 172002
ttZ	8 TeV	JHEP 01 (2016) 096
ttZ	13 TeV	JHEP 03 (2020) 056
ty	13 TeV	PRL 121 221802 (2018)
ttW	8 TeV	JHEP 01 (2016) 096
ttW	13 TeV	JHEP 07 (2023) 219
tWZ	13 TeV	TOP-22-008
tttt	13 TeV	Submitted to PLB

ggH	7 TeV	EPJC 75 (2015) 212
ggH	8 TeV	EPJC 75 (2015) 212
ggH	13 TeV	Nature 607 60-68 (2022)
VBF qqH	7 TeV	EPJC 75 (2015) 212
VBF qqH	8 TeV	EPJC 75 (2015) 212
VBF qqH	13 TeV	Nature 607 60-68 (2022)
VH	8 TeV	EPJC 75 (2015) 212
WH	13 TeV	Nature 607 60-68 (2022)
ZH	13 TeV	Nature 607 60-68 (2022)
ttH	8 TeV	EPJC 75 (2015) 212
ttH	13 TeV	Nature 607 60-68 (2022)
tH	13 TeV	Nature 607 60-68 (2022)
HH	13 TeV	Nature 607 60-68 (2022)



**LHC experiments  
 need accurate parton  
 distribution functions  
 (PDFs)**

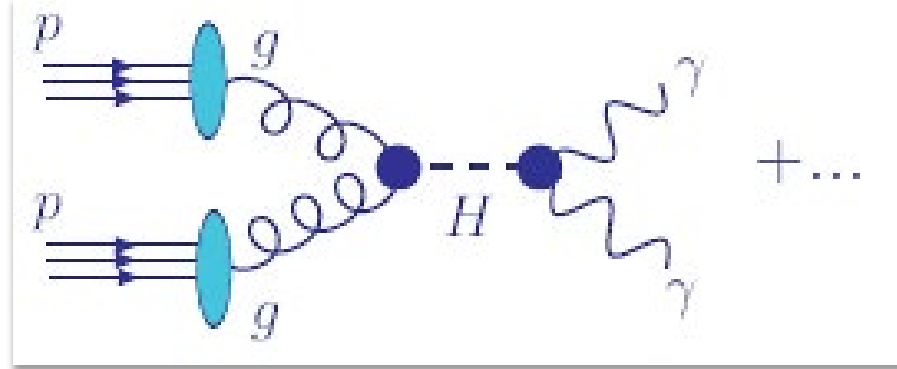
Measured cross sections and exclusion limits at 95% C.L.

See here for all cross section summary plots

2024-05-24

P. Nadolsky, TAMU

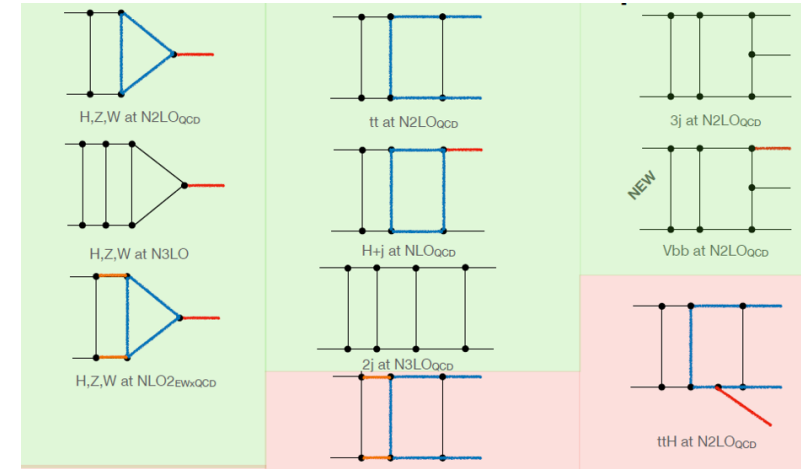
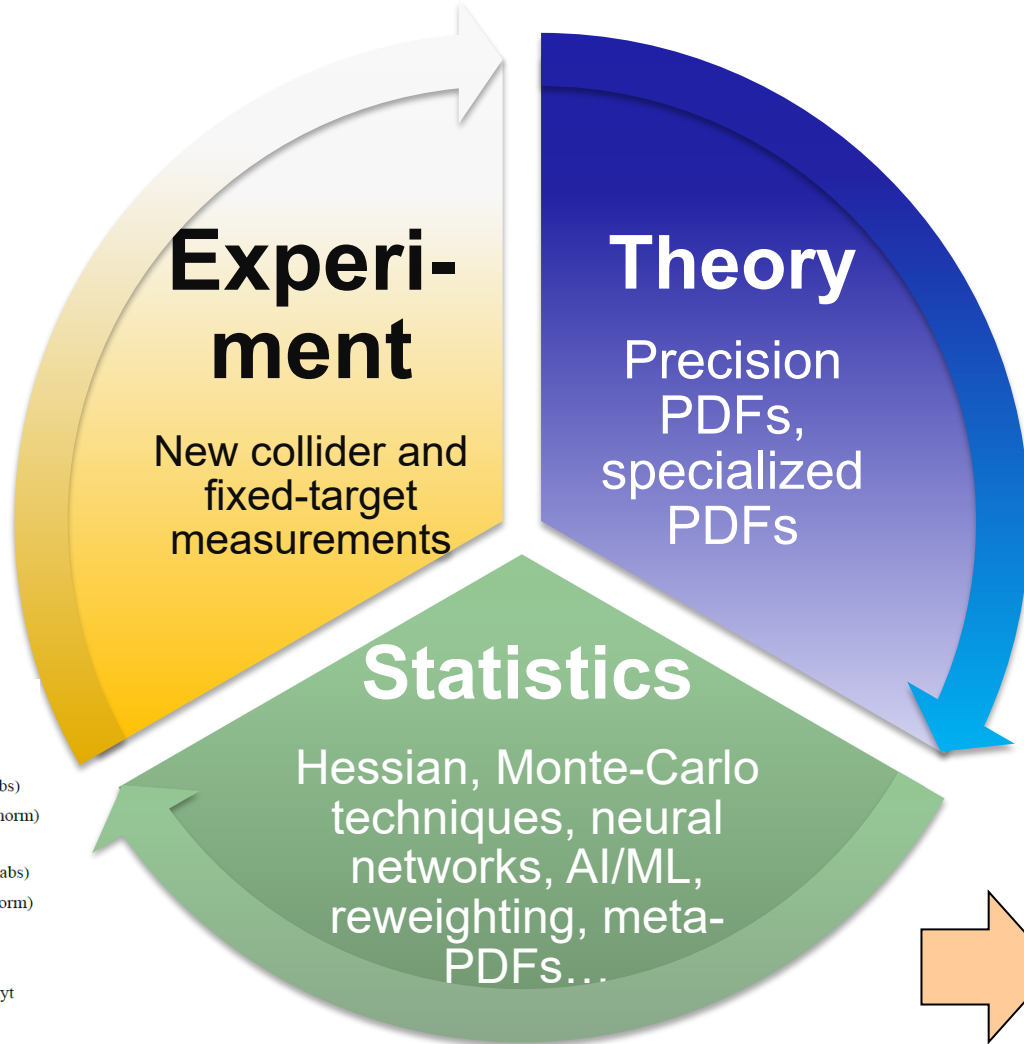
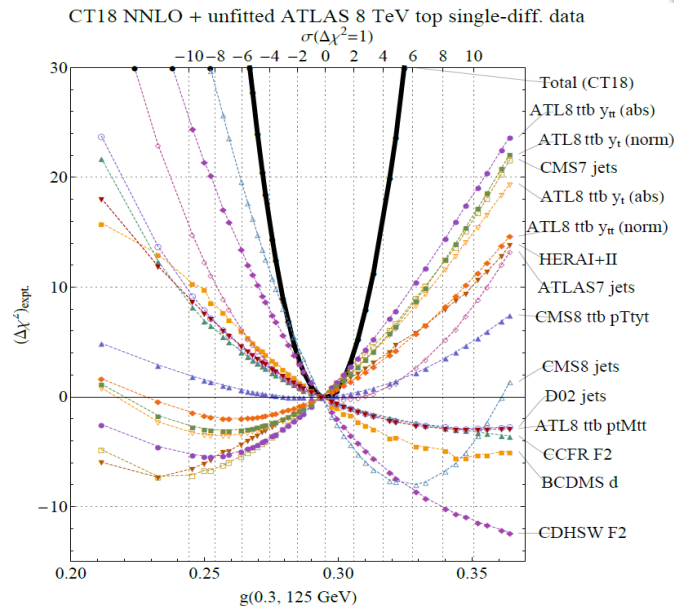
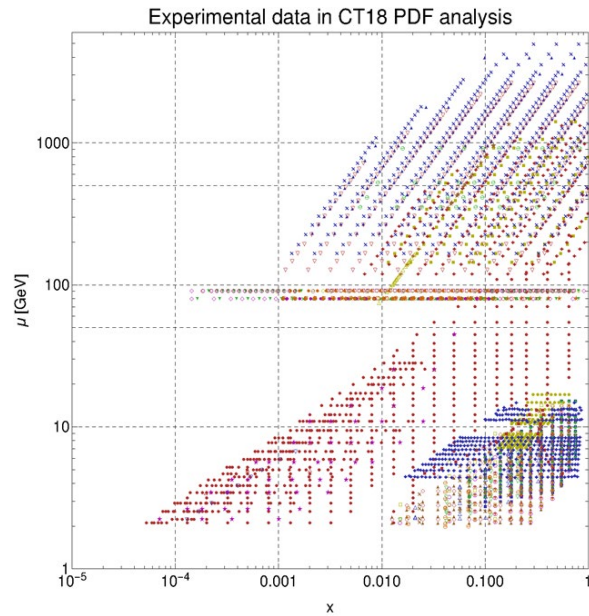
## Parton distributions describe long-distance dynamics in high-energy collisions



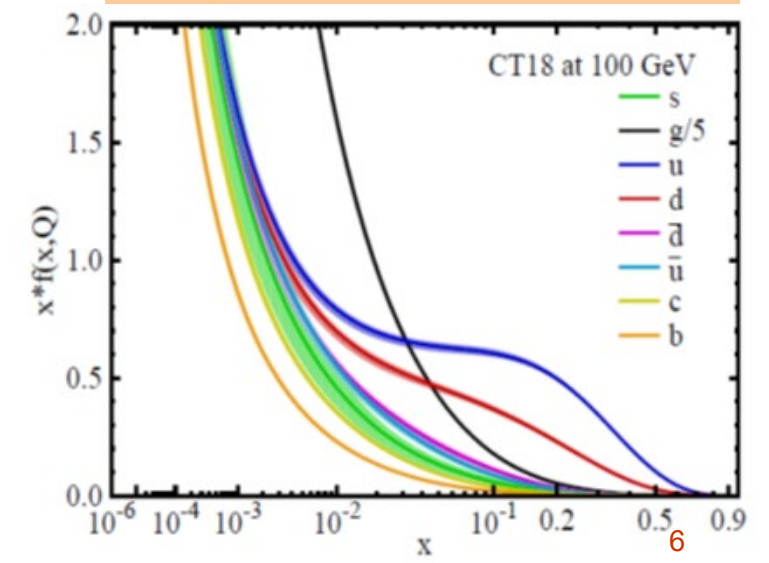
$$\sigma_{pp \rightarrow H \rightarrow \gamma\gamma X}(Q) = \sum_{a,b=g,q,\bar{q}} \int_0^1 d\xi_a \int_0^1 d\xi_b \hat{\sigma}_{ab \rightarrow H \rightarrow \gamma\gamma} \left( \frac{x_a}{\xi_a}, \frac{x_b}{\xi_b}, \frac{Q}{\mu_R}, \frac{Q}{\mu_F}; \alpha_s(\mu_R) \right) \\ \times f_a(\xi_a, \mu_F) f_b(\xi_b, \mu_F) + O\left(\frac{\Lambda_{QCD}^2}{Q^2}\right)$$

$\hat{\sigma}$  is the hard cross section; computed order-by-order in  $\alpha_s(\mu_R)$   
 **$f_a(x, \mu_F)$  is the distribution for parton  $a$  with momentum fraction  $x$ , at scale  $\mu_F$**

# Global fits of proton scattering data at (N)NNLO accuracy



## Parton distribution functions with uncertainties



# Global fits of proton scattering data at (N)NNLO accuracy

A rich domain of SM phenomenology!

Impact on a wide range of HEP and NP studies

Multiloop QCD and EW computations

Exploration of most complex experimental data sets

Accurate and fast high-performance computing

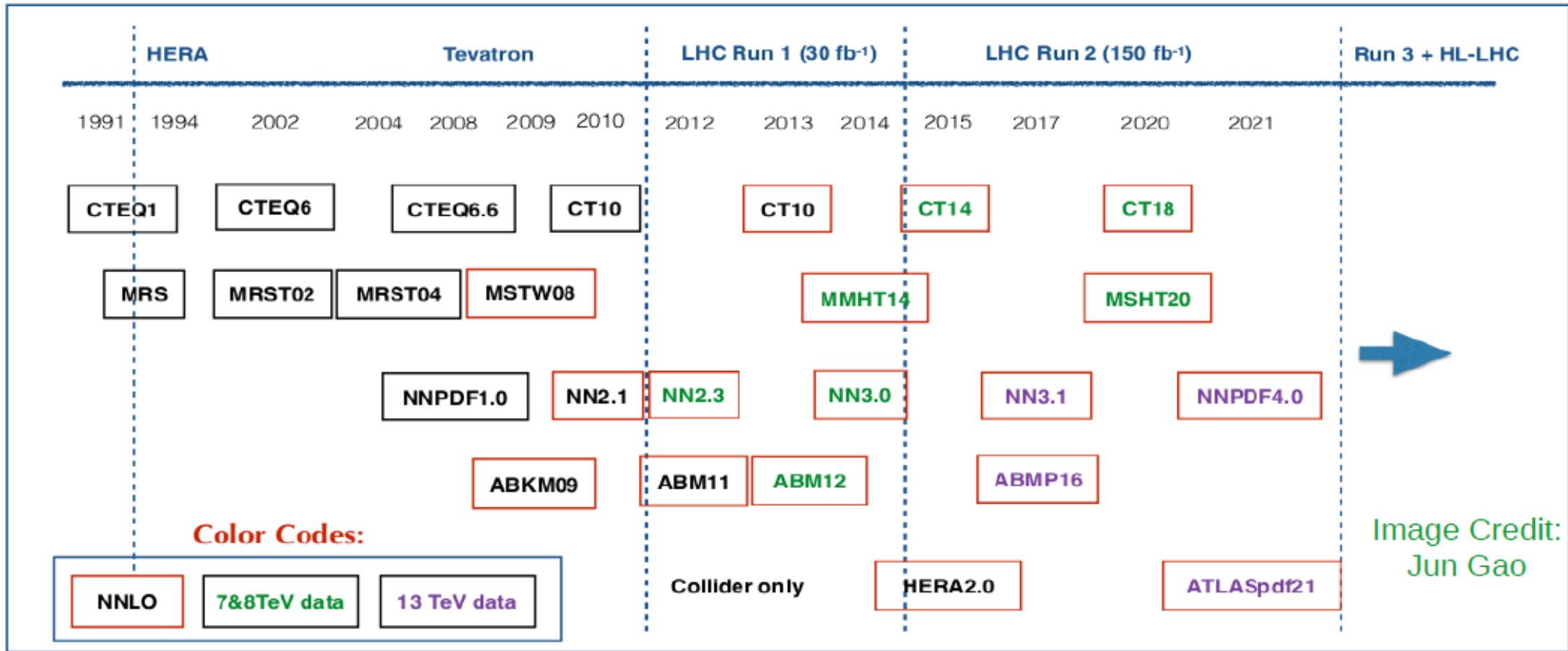
Frontier statistical inference in many dimensions

A testing bed for multidimensional uncertainty quantification, ML/AI, ...



opportunities  
for conceptual  
breakthroughs

# Phenomenological PDF analyses for a nucleon



Pursued by several groups – **ABM**, **ATLAS**, **CTEQ-TEA (CT)**, **CTEQ-Jlab**, **MSHT**, **NNPDF**, **JAM**, ...

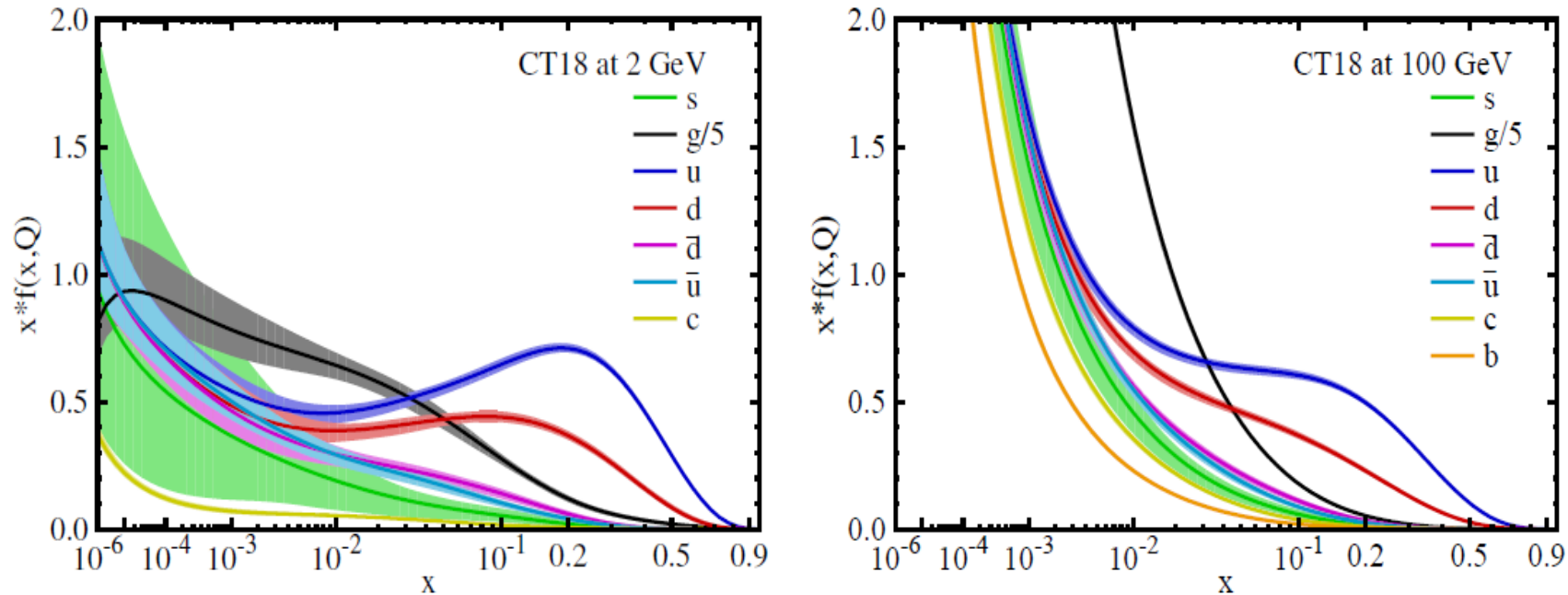
Precision state-of-the art: **NNLO QCD + NLO EW**; partial **N3LO** results (**NNPDF** and **MSHT** groups)

Data from fixed-target experiments and colliders (HERA, Tevatron, LHC, ...)



# CT18 parton distributions

Recent PDFs from the CTEQ-TEA group arXiv:[1912.10053](https://arxiv.org/abs/1912.10053) [hep-ph]



- Precise experimental data sets from  $ep$  collider HERA, LHC, Tevatron, fixed-target experiments
- Next-to-next-to-leading order (NNLO) accuracy in QCD coupling  $\alpha_s$
- Flexible parametric forms
- Central PDFs and bands of estimated uncertainty
- Four PDF ensembles, to account for tensions between data sets

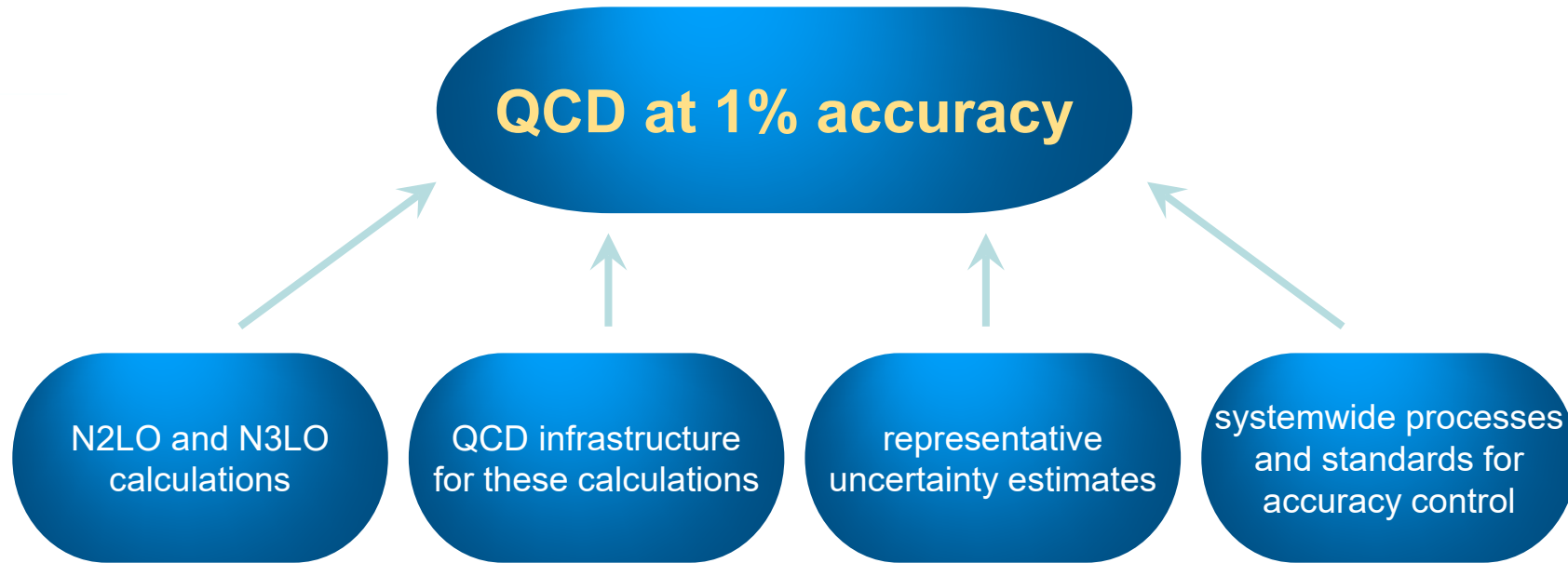
# Toward a new generation of CT202X PDFs

1. Multiple preliminary NNLO fits with LHC Run-2 (di)jet, vector boson,  $t\bar{t}$  data
  - based on the selections of experiments recommended in 2305.10733, 2307.11153
2. Work on implementation of N3LO contributions
3. Physics applications
  - a. QCD+QED PDFs for a proton and neutron
  - b. PDF dependence of forward-backward asymmetry
  - c. Fast tools for PDF profiling and examination of data tensions
  - d. Pion PDFs
  - e. ...
4. Next-generation PDF uncertainty quantification: Bézier curves, META combination, ML stress-testing, multi-Gaussian approaches, ...

# Multivariate uncertainty quantification in the global PDF fits

- More than 20 years of experience on UQ by CTEQ and other groups
- Essential for statistical analyses of numerous experiments
- Understanding of mathematical foundations for UQ in multivariate and AI frameworks
- Provides guidelines for replicability of precision HEP analyses

**Precision**  
**Reproducibility & replicability**  
**Epistemic uncertainty**  
**Big-data paradox**  
**AI applications**



Lots of promise in this area

Parton showers, fast NxLO interfaces, PDFs, ... must be comparably accurate

or The Importance of Being Earnest with Systematic Errors (experiment+theory; traditional or AI/ML)

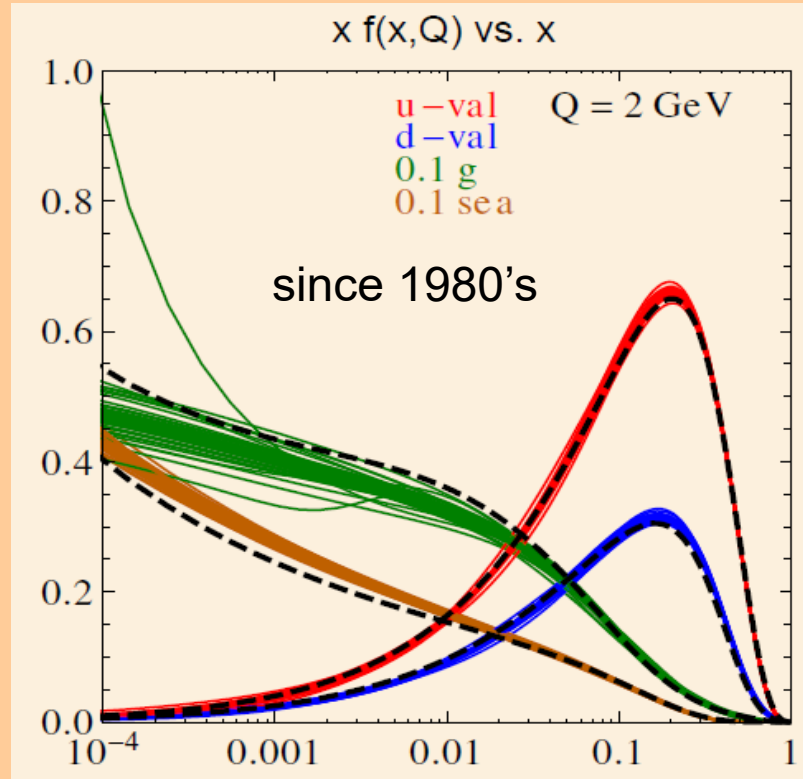
This must be a part of the precision-focused community culture

Publishing statistical models: Getting the most out of particle physics experiments  
 Kyle Cranmer (New York U.), Sabine Kraml (LPSC, Grenoble), Harrison B. Prosper (Florida State U.), Philip Bechtle (Bonn U.), Florian U. Bernlochner (Bonn U.) [Show All\(33\)](#)  
 Sep 10, 2021  
 60 pages  
 Published in: *SciPost Phys.* 12 (2022) 1, 037, *SciPost Phys.* 12 (2022) 037  
 Published: Jan 25, 2022  
 e-Print: [2109.04981](#) [hep-ph]

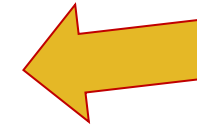
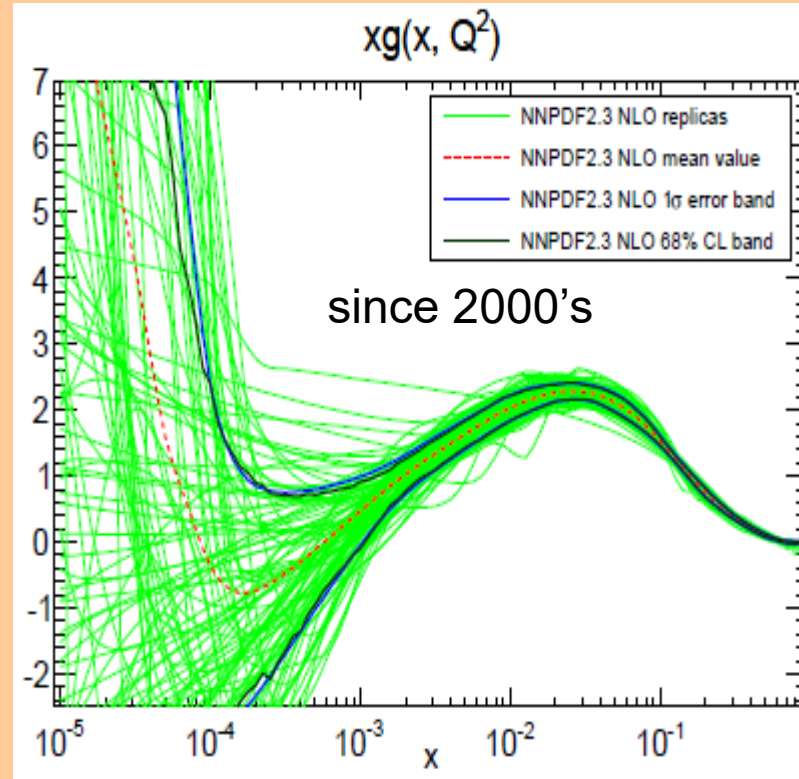
2023 US DOE Funding Opportunity Announcement  
 DE-FOA-0000315  
**Advancing *Uncertainty Quantification* in Modeling, Simulation, and Analysis of *Complex Systems***

# Two complementary approaches to estimate PDF uncertainties

Analytic parametrizations +  
Hessian PDF eigenvector sets  
**(ABM, CTEQ, HERA, MSHT,...)**



Neural network parameterizations  
+ Monte Carlo PDF replicas  
**(NNPDF)**

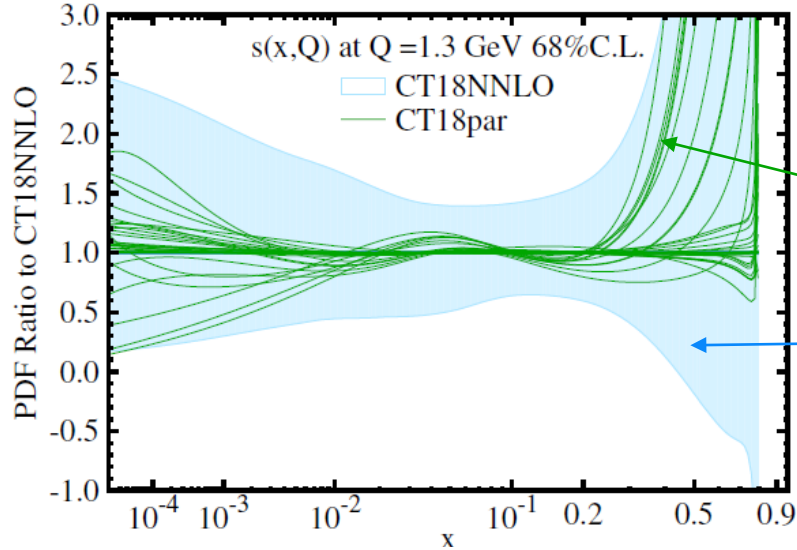


a textbook  
application of  
ML in particle  
physics

Hessian PDFs can be converted into MC ones, and vice versa.

# Hessian PDFs, uncertainties (CT18 PDFs)

“Bayesian exploration with Gaussian emulation”



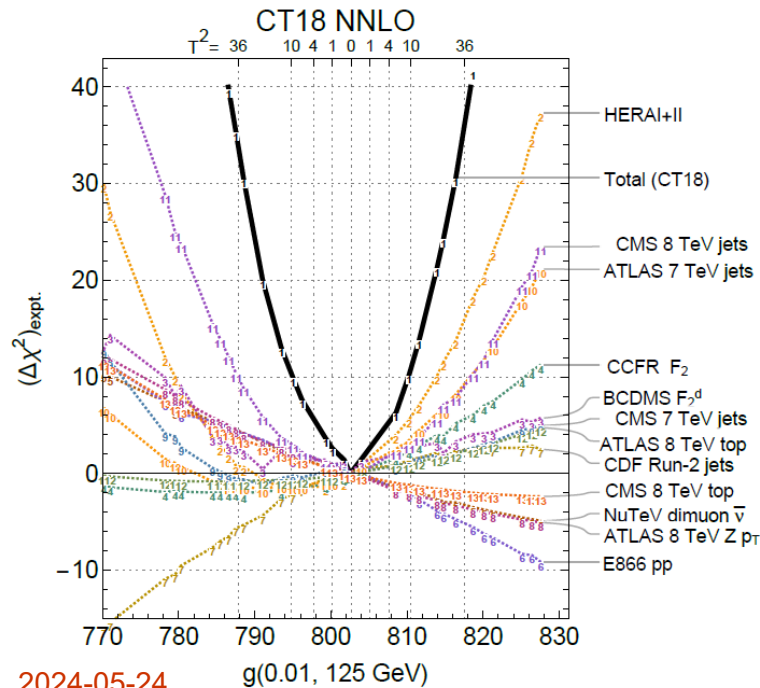
preliminary PDFs for alternative parametrizations

final uncertainty with one parametrization

Preliminary fits explore experimental, theoretical, parametrization, methodological uncertainties

The final PDFs are released as one quasi-Gaussian (**Hessian**) error set (50-60) that approximates the total uncertainty due to the above factors.

These error sets are constructed with a fixed choice of polynomial parametrization forms. The totality of error sources (not only experimental) is emulated by introducing **tolerance**  $T$ : the final  $1\sigma$  uncertainty corresponds to  $\Delta\chi^2 = T^2 \sim 10 - 30$ , rather than  $T^2 = 1$ .



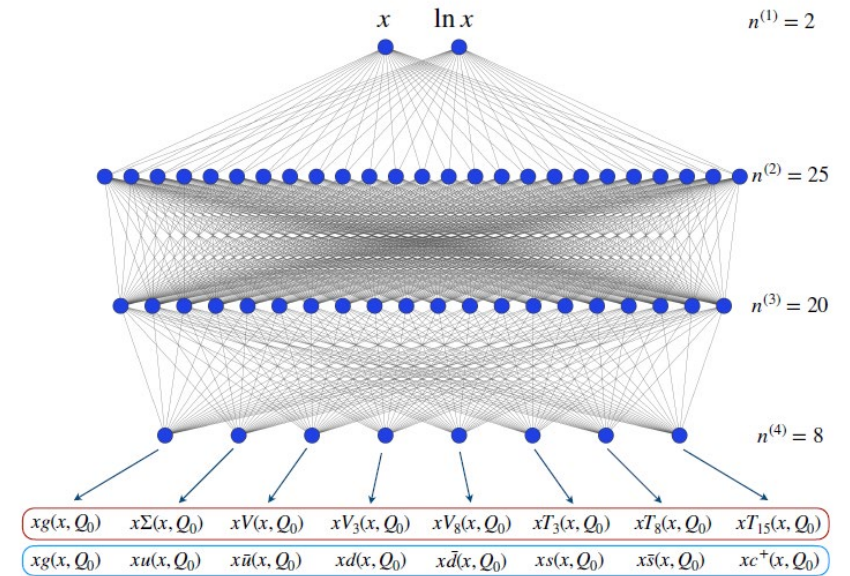
# An alternative: Neural-network PDFs

Use **bootstrap** to estimate **aleatory** data fluctuations for a fixed training methodology (called “importance sampling” by NNPDF)

Parametrize PDFs using CNNs with optimized hyperparameters and restricted by prior conditions (positivity of cross sections, etc.)

[The whole fit is based on one “tuned” NN architecture]

All fitting codes, especially the NNPDF one, employ grid techniques for fast integration



**Figure 3.9.** The neural network architecture adopted for NNPDF4.0. A single network is used, whose eight output values are the PDFs in the evolution (red) or the flavor basis (blue box). The architecture displayed corresponds to the optimal choice in the evolution basis; the optimal architecture in the flavor basis is different as indicated by Table 3.3).

NNPDF4.0 PDF ensemble,  
R. Ball et al., arXiv:2109.02653

# Statistics with many parameters is different!

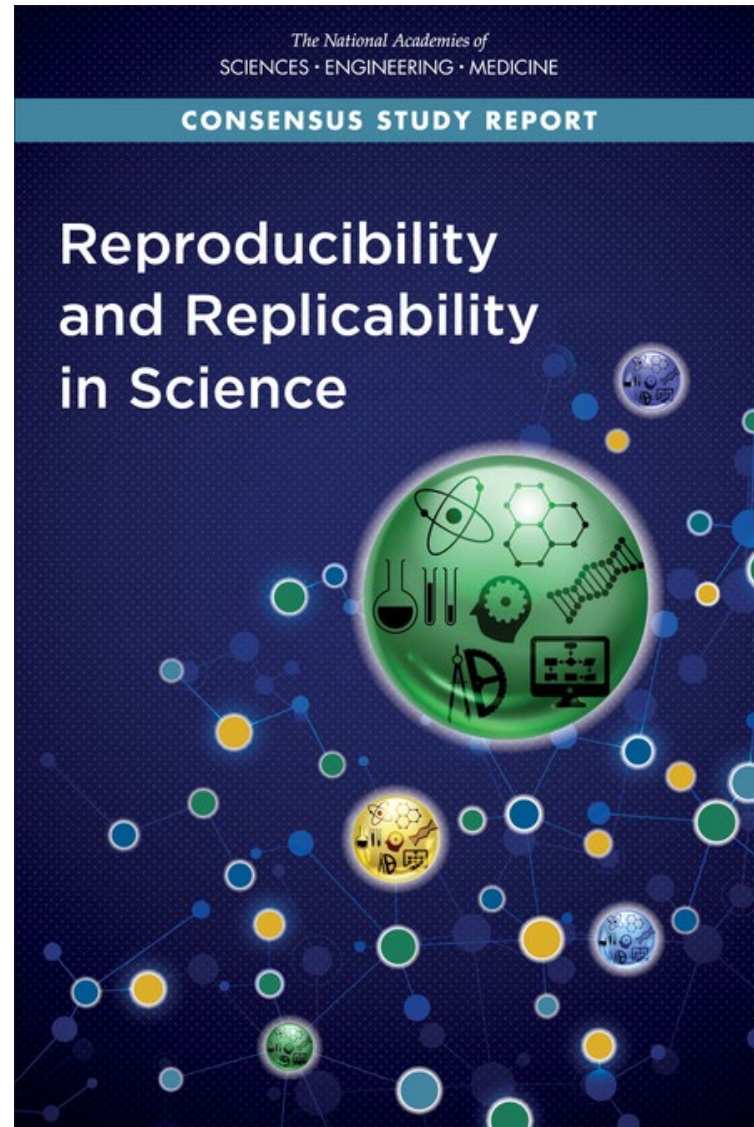
In many applications, especially AI/ML ones:

1. **There is no single global minimum of  $\chi^2$  (or another cost function)**
2. **The law of large numbers may not work**
  - uncertainty may not decrease as  $1/\sqrt{N_{\text{rep}}}$ , leading to the **big-data paradox** [Xiao-Li Meng, 2018]:

**The bigger the data, the surer we fool ourselves.**

3. **Replication of complex measurements is daunting**





*US National Academy of Sciences, Engineering, and Medicine, 2019, <https://doi.org/10.17226/25303>*

# Are $W$ boson mass measurements replicable?

For instance,  $W$  boson mass measurements at the Tevatron and LHC

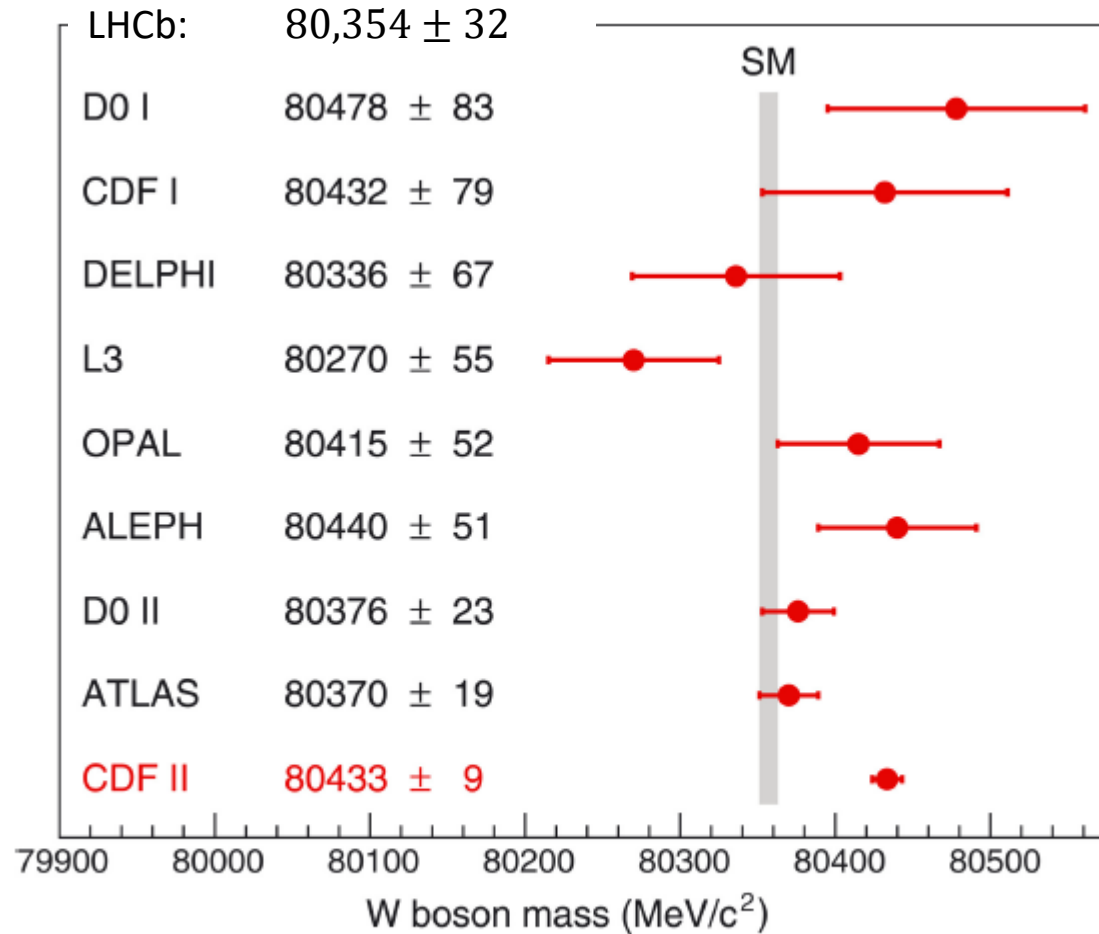
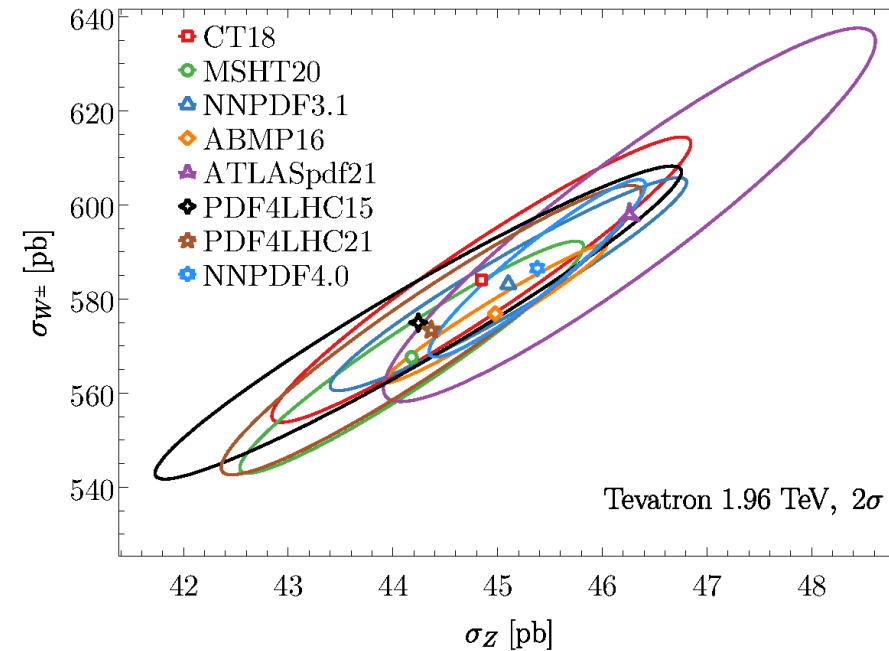
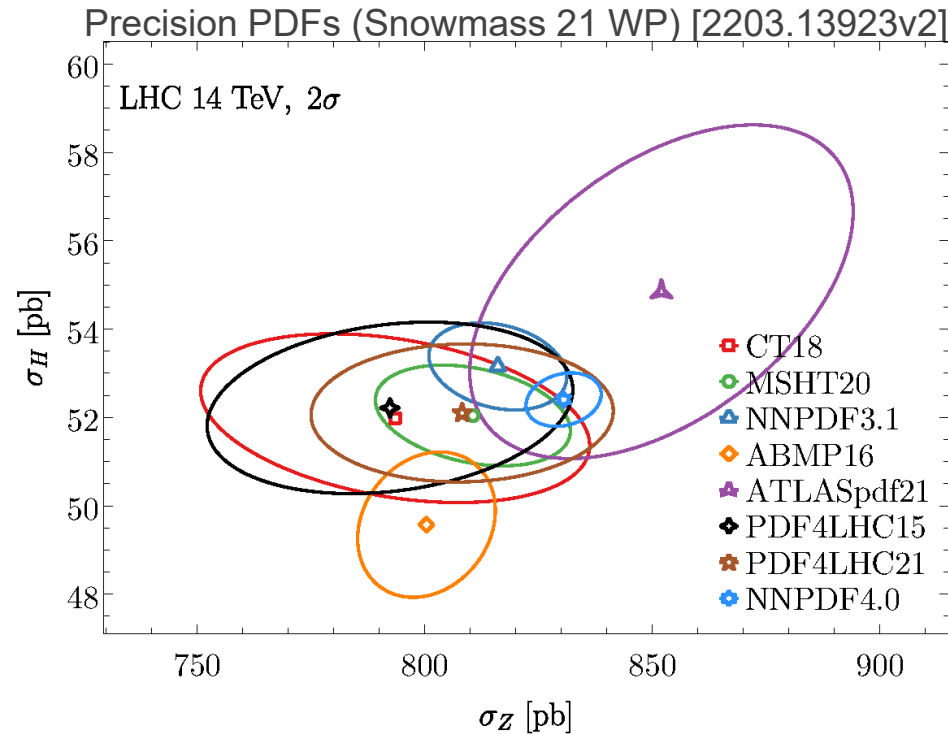


Figure reproduced from CDF-II measurement (Science 376, 170).

# The tolerance puzzle

Why do groups fitting similar data sets obtain different PDF uncertainties?



The answer has direct implications for high-stake experiments such as  $W$  boson mass measurement, tests of nonperturbative QCD models and lattice QCD, high-mass BSM searches, etc.

## Comparisons of the latest PDF sets

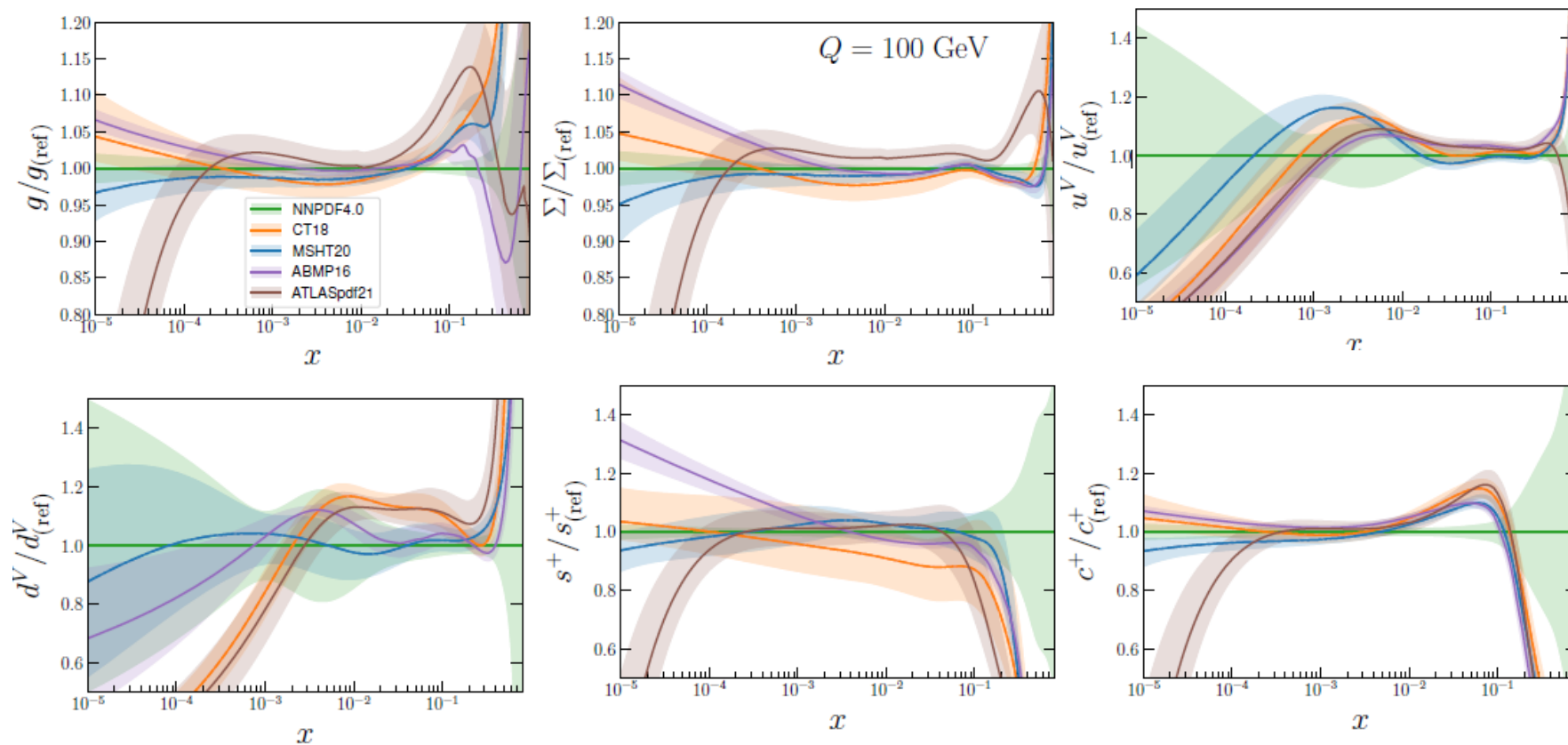


FIG. 2. Comparison of the PDFs at  $Q = 100$  GeV. The PDFs shown are the N2LO sets of NNPDF4.0, CT18, MSHT20, ABMP16 with  $\alpha_s(M_Z) = 0.118$ , and ATLASpdf21. The ratio to the NNPDF4.0 central value and the relative  $1\sigma$  uncertainty are shown for the gluon  $g$ , singlet  $\Sigma$ , total strangeness  $s^+ = s + \bar{s}$ , total charm  $c^+ = c + \bar{c}$ , up valence  $u^V$  and down valence  $d^V$  PDFs.

# Replicability risks for precision HEP

Nearly all complex STEM fields encounter replicability challenges.

Modern particle physics is not an exception.

1. It is complex! Is it rigorous enough?
  - Many approaches, especially AI-based ones, increase complexity and are not rigorously understood
2. It often uses wrong prescriptions for estimating epistemic uncertainties
  - Tens to hundreds of systematic uncertainties affect measurements, phenomenology, and lattice QCD

# Ongoing studies of systematic uncertainties are essential and still insufficient

- from the experiment side

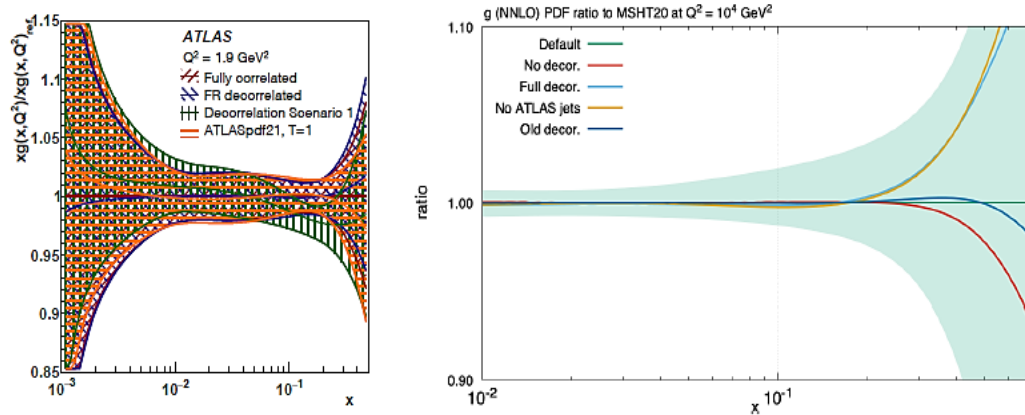


FIG. 9. Difference in the gluon PDF shown in ratio to the ATLASpdf21 (default) gluon (left). This default uses Decorrelation Scenario 2 and this is compared to the use of Full Correlation, Full decorrelation of the flavour response systematic and Decorrelation Scenario 1. The effect of no decorrelation, the default correlation of [9], the decorrelation in [362], and full decorrelation for the MSHT20 gluon (right).

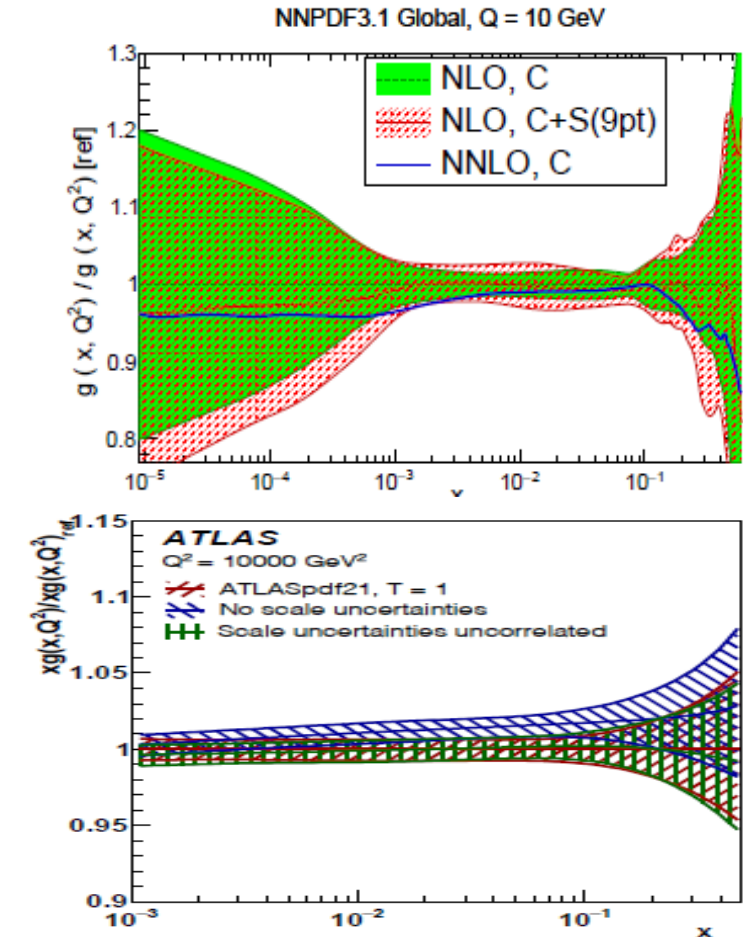
*S. Amoroso et al., 2203.13923, Sec. 5.A*

Strong dependence on the definition of corr. syst. errors raises a general concern:

**Overreliance on Gaussian distributions and covariance matrices for poorly understood effects may produce very wrong uncertainty estimates**

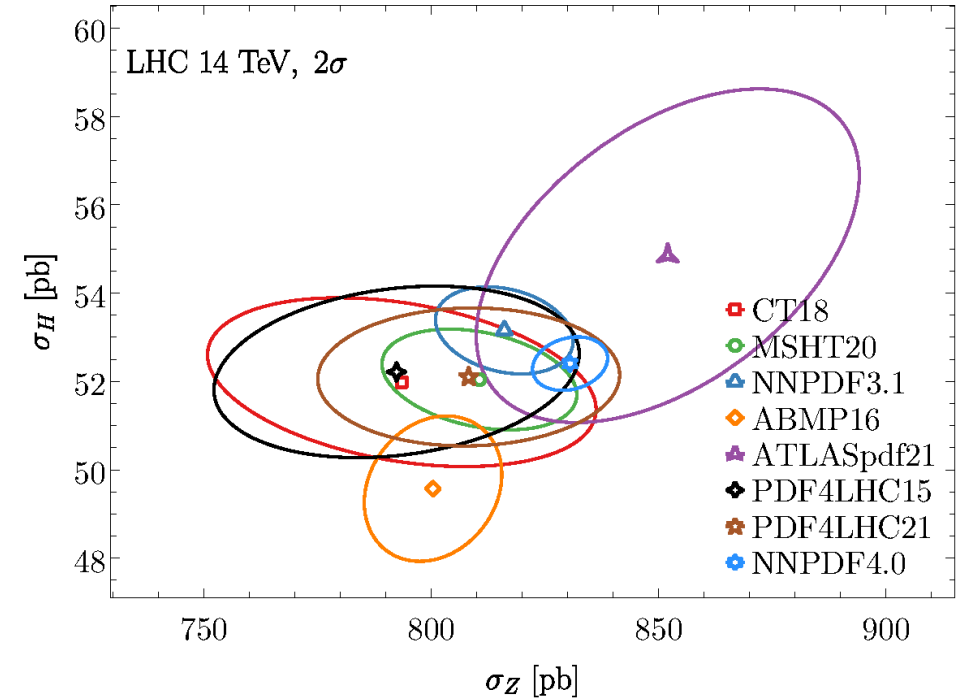
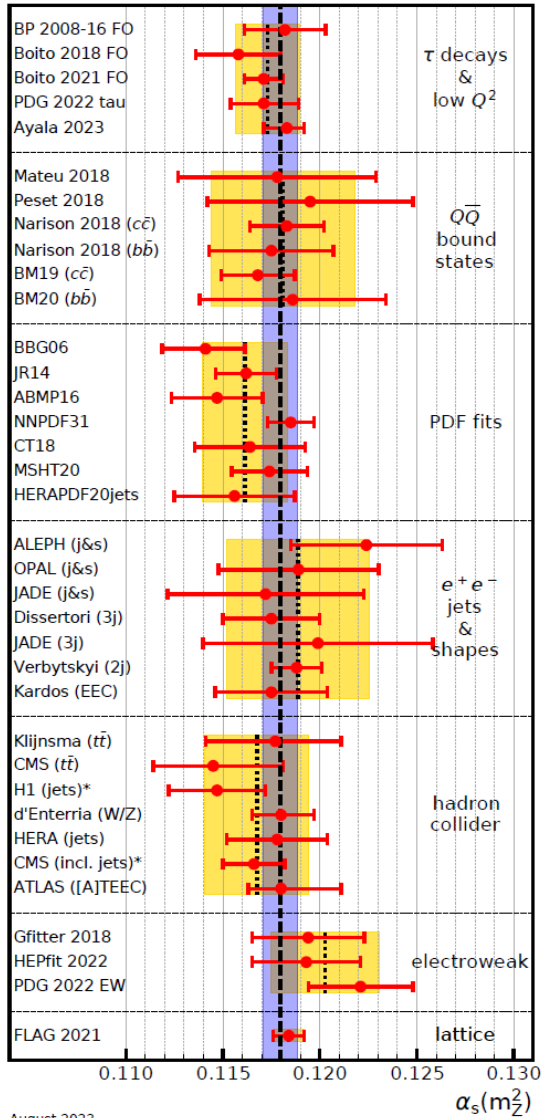
[N. Taleb, Black Swan & Antifragile]

- from the theory side



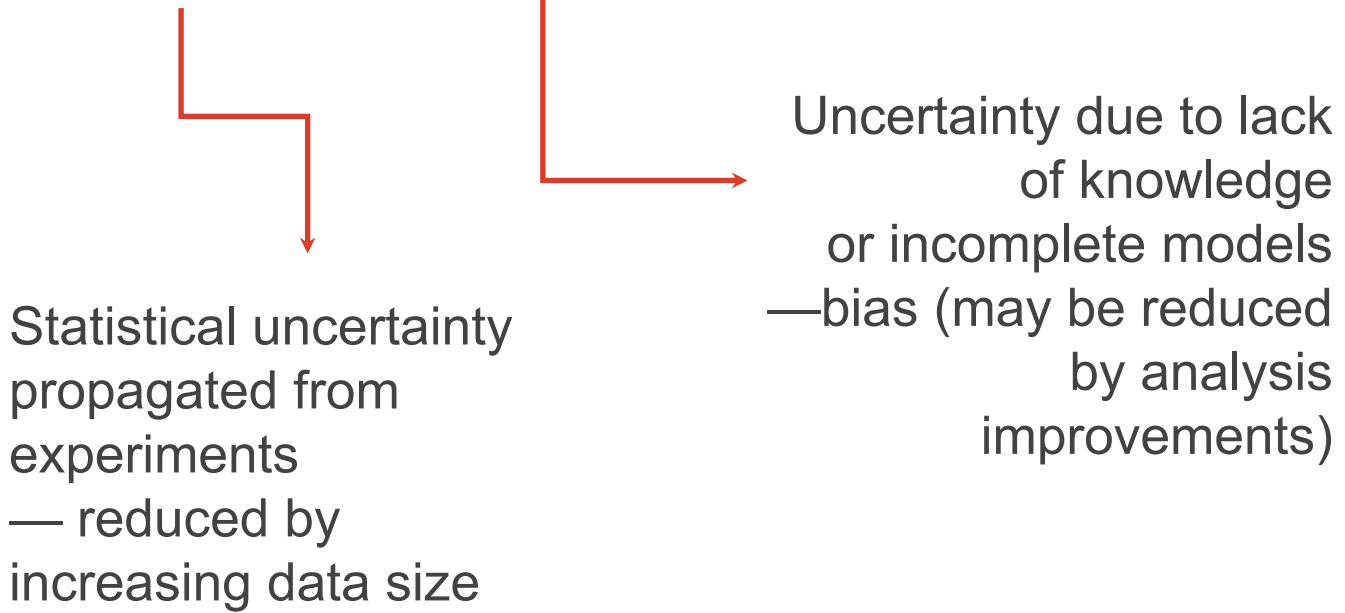
Examples: studies of theory uncertainties in the PDFs by NNPDF3.1 and ATLAS21

# Replicability and PDF uncertainties

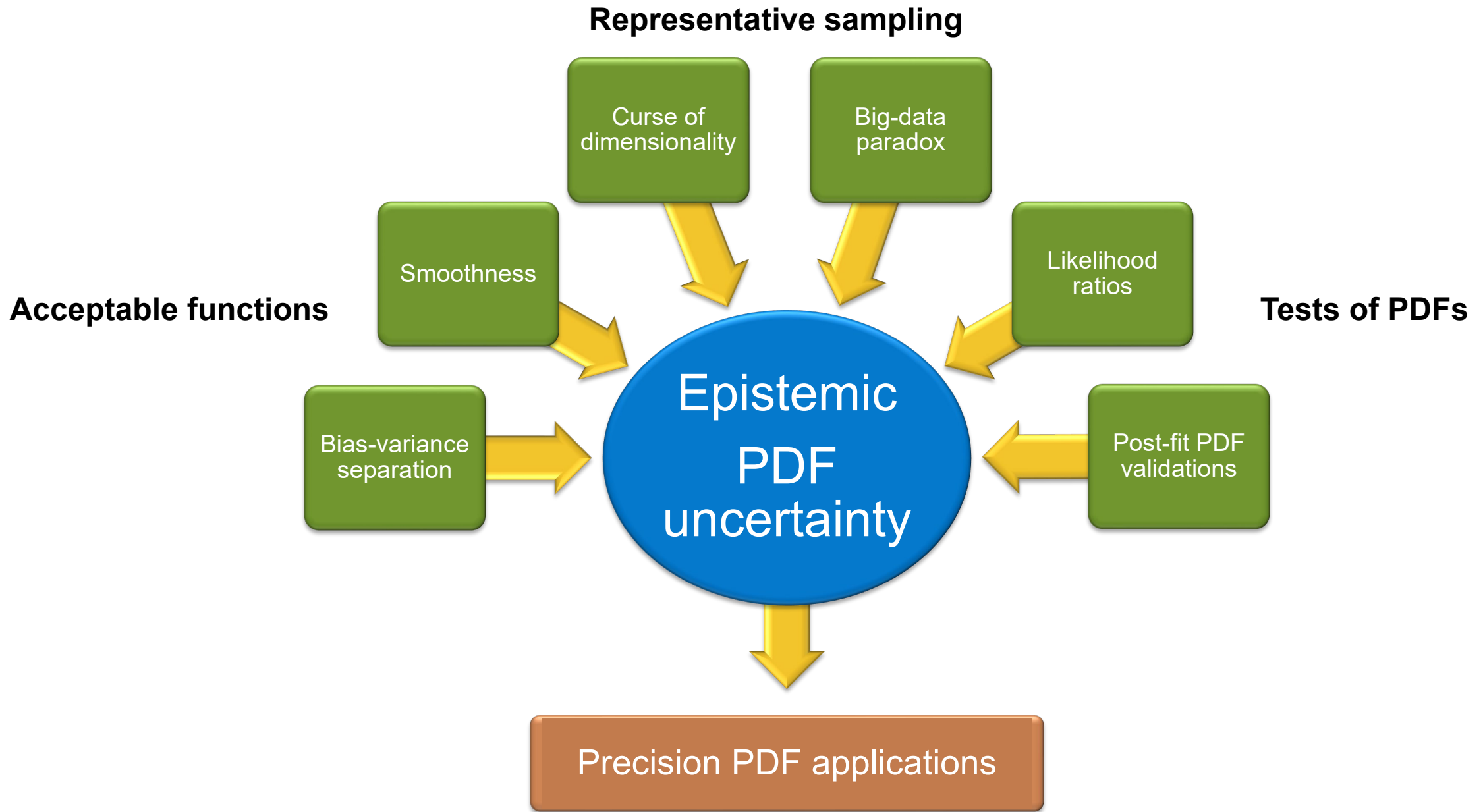


Quantification of **epistemic** PDF uncertainties is a central factor affecting **replicability** of upcoming determinations of the QCD coupling constant  $\alpha_s$ , Higgs couplings, mass of weak bosons.

## aleatory vs. epistemic uncertainties







# Epistemic PDF uncertainty...

...reflects **methodological choices** such as PDF functional forms, NN architecture and hyperparameters, or model for systematic uncertainties

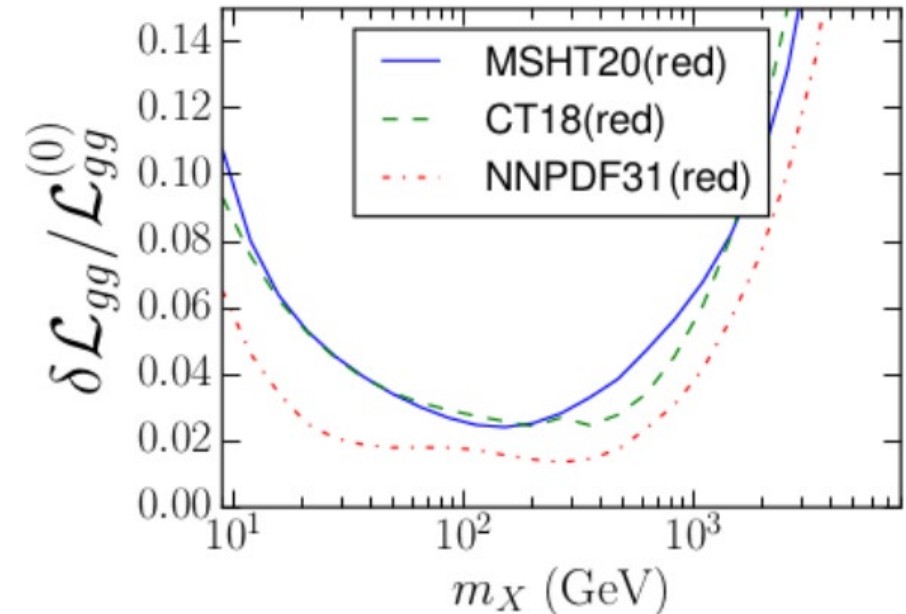
... can dominate the full uncertainty when experimental and theoretical uncertainties are small.

...is associated with the **prior probability**.

... can be estimated by **representative sampling** of the PDF solutions obtained with acceptable methodologies.

⇒ sampling over choices of experiments, PDF/NN functional space, models of correlated uncertainties...

⇒ in addition to sampling over data fluctuations

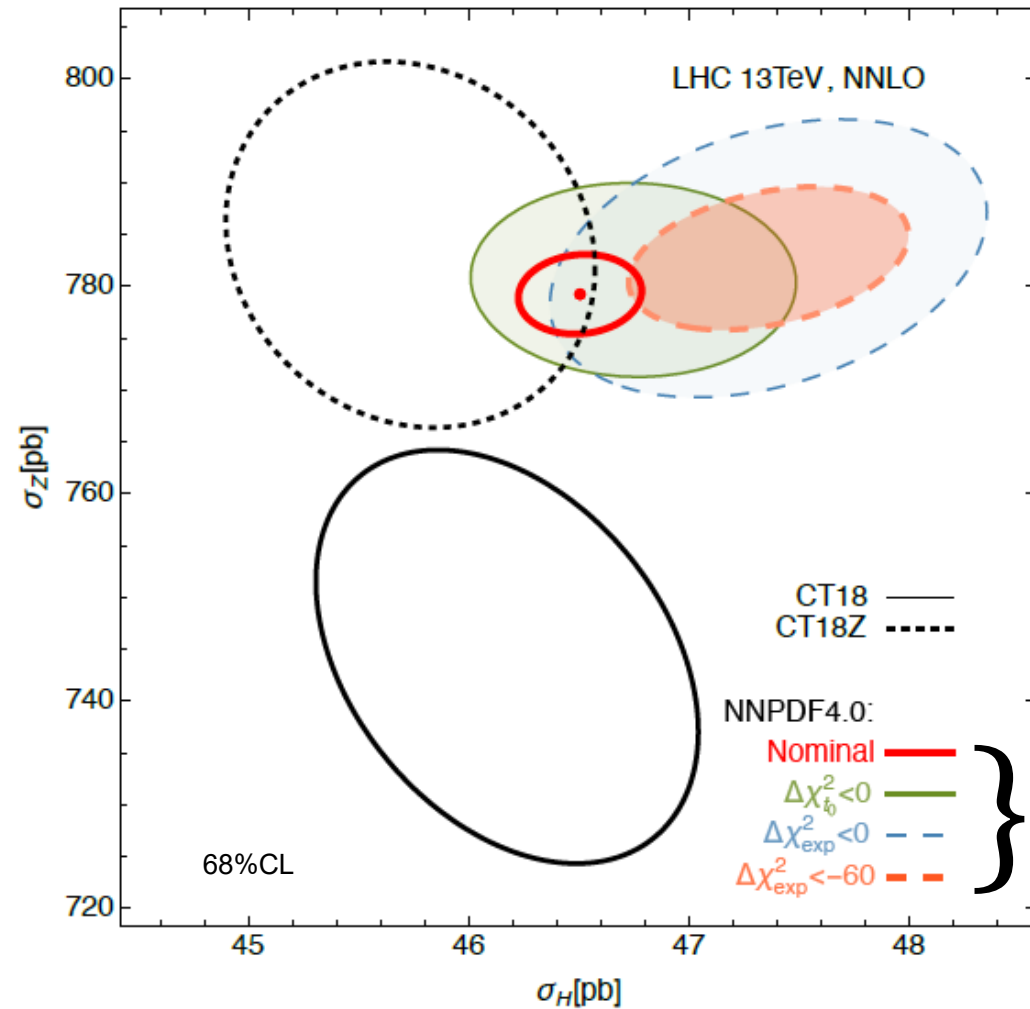


Epistemic uncertainties explain many of the differences among the sizes of PDF uncertainties by CT, MSHT, and NNPDF global fits to the same or similar data

Details in [arXiv:2203.05506](https://arxiv.org/abs/2203.05506), [arXiv:2205.10444](https://arxiv.org/abs/2205.10444)

# Example: the impact of epistemic uncertainty on NNLO Higgs and Z cross sections

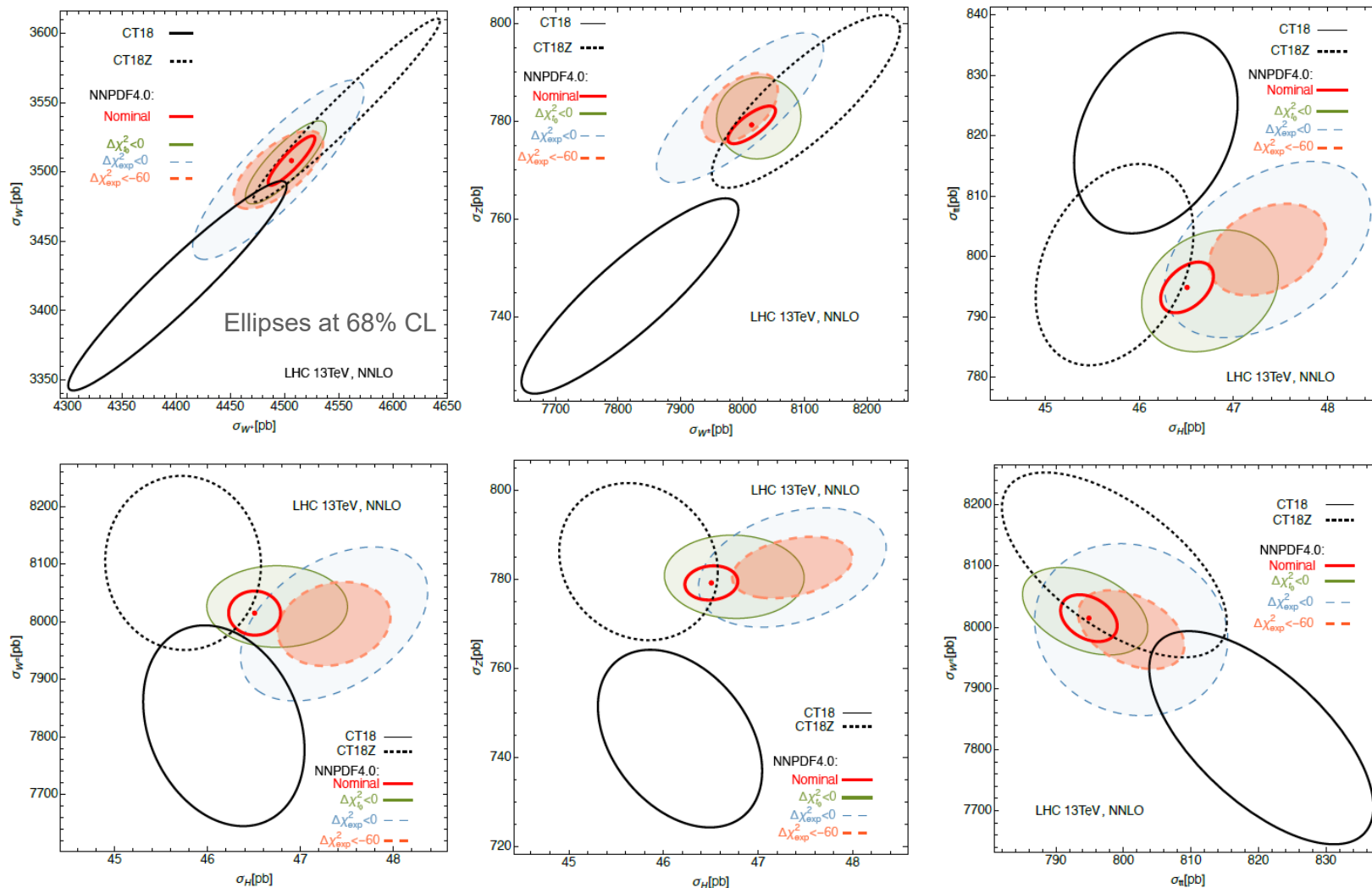
Details in  
A. Courtoy et al.,  
[arXiv:2205.10444](https://arxiv.org/abs/2205.10444)



obtained with the same NNPDF4.0 fitting code  
using a “**hopscotch scan**” of the PDF param. space

all ellipses contain acceptable predictions  
according to the likelihood-ratio test  
Nominal NN4.0 uncertainty is too small!

# Impact of epistemic uncertainties on other cross sections

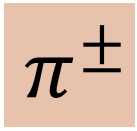


The ellipses are projections of 68% c.l. ellipsoids in  $N_{par}$ -dim. spaces

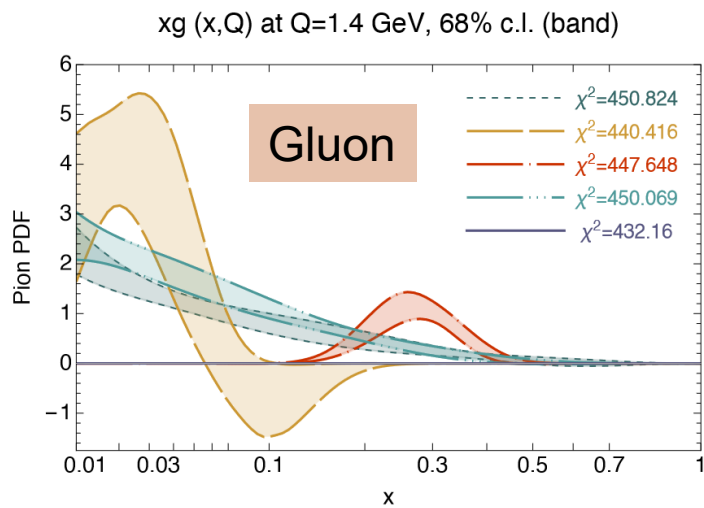
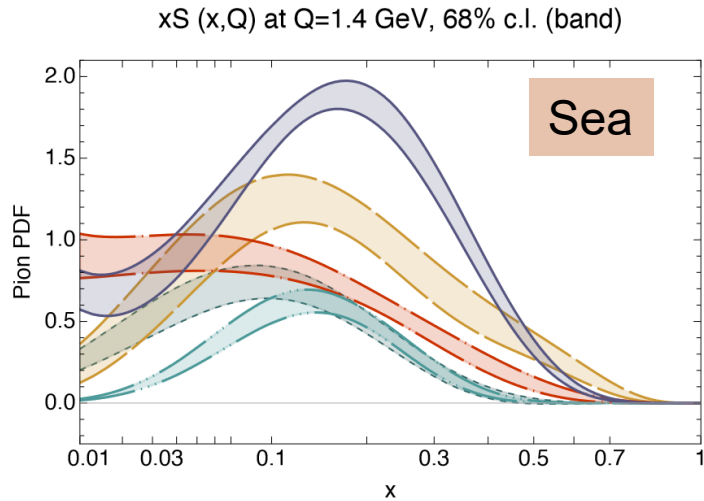
$N_{par} = 28$  and  $50$  for CT18 and NNPDF4.0 Hessian PDFs

# New approaches to determine the PDF tolerance

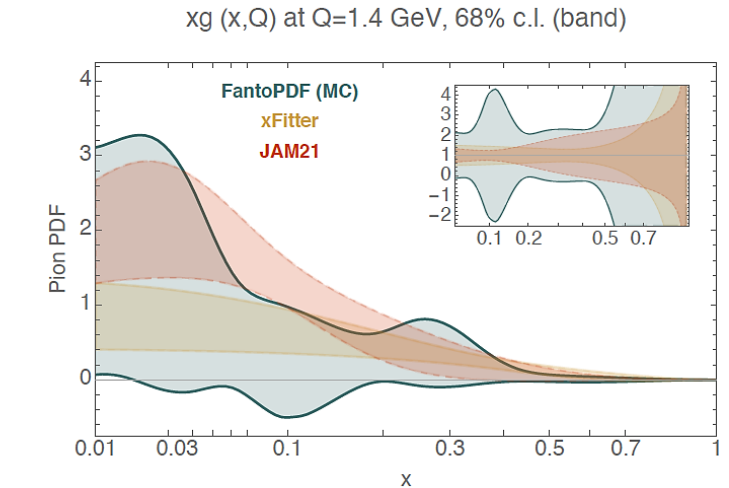
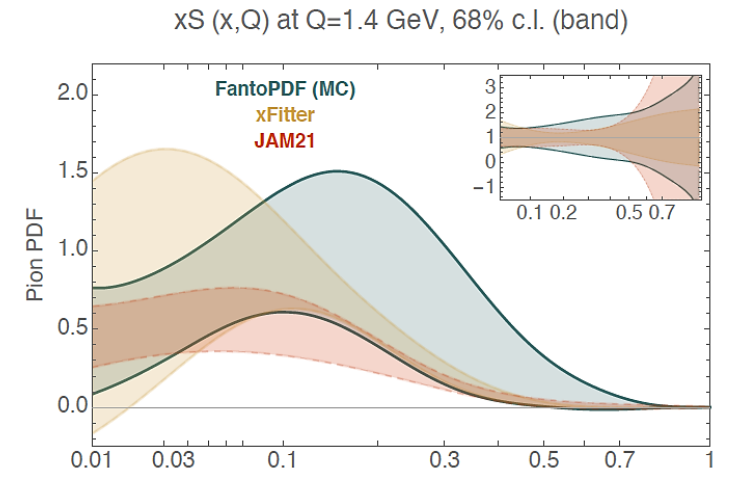
## On the example of Fantômas pion PDFs



L. Kotz, A. Courtoy, M. Chavez, P. N., F. Olness, arXiv:2311.08447



META PDF technique  
combines aleatory error  
bands for multiple  
parametrization forms



## Final remarks

*Epistemic uncertainty* (due to parametrization, methodology, parametrization/NN architecture, smoothness, data tensions, model for syst. errors, ...) is increasingly important in (N)NNLO global fits as experimental and theoretical uncertainties decrease

Nominal PDF uncertainties in high-stake measurements (ATLAS W mass, Higgs cross sections...) thus should be tested for *control of tensions* and *robustness of sampling over acceptable methodologies*.

*Smoothness of Hessian and NN PDFs* is another such aspect associated with the prior that should be explored.

Tools for such studies already exist, e.g. *hopscotch scans*.

This is also necessary for combination of PDFs including data correlations

[LHC EW, Jet & Vector boson WGs, <https://tinyurl.com/4wcnd8xn>; <https://tinyurl.com/2p8d8ba3>; <https://tinyurl.com/2p8tcn5b>; Ball, Forte, Stegeman, arXiv:[2110.08274](https://arxiv.org/abs/2110.08274)].

Better control of the epistemic uncertainties is central for *replicability* of upcoming QCD measurements.

# Strategies for improving replicability and reproducibility

Preselection of planned studies based on their likely replicability

Detailed documentation of methods and uncertainty quantification in the publications

Training of researchers in relevant statistical methods

Journal policies that encourage replicability

Support from the funding agencies for the research infrastructure and collaborations focusing on replicability

Support for open publication of the analysis codes and key data, using agreed-upon formats

“Skin-in-the-game” incentives for researchers to produce replicable results

*Based on “REPRODUCIBILITY AND REPLICABILITY IN SCIENCE”*